## Contents

Chapter I. Background ........................................... 1  
  1. Terminology .................................................. 1  
  2. Calculus Review ............................................. 2  
  3. Complex Numbers Review ................................... 3  
  4. Worked-Out Examples ....................................... 6  

Chapter II. First-Order, Differential Equations ............... 11  
  1. Basics ......................................................... 11  
  2. Linear Equations ............................................ 14  
  3. Separable and Autonomous Equations ..................... 16  
  4. Exact Equations ............................................. 18  
  5. Worked-Out Examples ....................................... 20  

Chapter III. Discrete Problems ................................ 29  
  1. First-order difference equations ......................... 29  
  2. Euler's Method .............................................. 29  
  3. Worked-Out Examples ....................................... 32  

Chapter IV. Second-Order, Linear, Differential Equations .... 37  
  1. Homogeneous Equations ..................................... 37  
  2. Non-Homogeneous Equations ............................... 40  
  3. Worked-Out Examples ....................................... 44  

Chapter V. Using Linear Algebra ................................. 65  
  1. Gaussian Elimination ....................................... 65  
  2. Matrices ...................................................... 69  
  3. Linear Algebra .............................................. 70  
  4. Worked-Out Examples ....................................... 74  

Chapter VI. Linear Systems of Differential Equations ......... 83  
  1. Homogeneous Systems with Constant Coefficients .... 84  
  2. Non-Homogeneous Systems ................................. 87  
  3. Linearization ............................................... 88  
  4. Worked-Out Examples ....................................... 91  

Appendix A. Unnecessary Proofs .................................. 107
Chapter I. Background

Section I.1. Terminology

It’s very important to know some of the common words used in talking about mathematical problems, and equations in particular. Many students think of an equation merely as something to be solved, or the start of a pre-determined procedure. More than a crutch to get through the class, this conception fundamentally cripples their ability to learn more advanced topics.

To start at the very basics, an equation is a statement. Like any statement, it can be either true or false. For example,

1. the equation $y = 5$ is true when $y$ is 5, but is false when $y$ is 4;
2. the equation $2y + 1 = 1$ is only true when $y = 0$;
3. the equation $2y + y = 3y$ is true for all $y$;
4. the equation $(y - 1)(y + 2) = 0$ is true for $y = 1$ and for $y = -2$;
5. the equation $y + 1 = y$ is never true.

To say that something solves an equation means that it makes the equation true. Similarly, something satisfies some condition or statement if that statement is true of it. Something witnesses that a statement is true, if that statement is true of it. These are all different words for the same thing. A solution is just something that solves the equation.

So, returning to (4) above, we might say that $y = 1$ is a solution to the equation, or that $y = -2$ witnesses that $(y - 1)(y + 2) = 0$. We might say that (5) above has no solution.

If we can describe all the possible solutions, this description is the general solution. Otherwise, we might only know a couple solutions, known as particular solutions. For example, $7x^3 + 2x^2 - 100x = 0$ has a particular solution $x = 0$, and the general solution is hard to find. $(x - 1)(x + 1) = 0$ has a general solution of $x = \pm 1$, where $x = 1$ and $x = -1$ are both particular solutions.

So solving an equation is just finding out when that equation is true. This is done through logic and reasoning. Many students, however, are not very good at this, and fail to understand that mathematics is about giving an argument, just as intelligible and legible as any other. This means complete sentences, and giving reasons. For example, consider the following line of flawed reasoning that attempts to show why $2y + y = 3y$ is always true:

1. $2y + y = 3y$.
2. Factor out a $y$: $(2 + 1)y = 3y$.
3. Divide by $y$: $2 + 1 = 3$.
4. Simplify: $3 = 3 \checkmark$

Such reasoning is mind-bogglingly wrong. Consider an analogous line of reasoning in plain English: I’m trying to show that all horses are animals.

1. All horses are animals.
2. For any horse, it is an animal.
3. Since all horses are animals, an animal is an animal, which we know is true.

Clearly this doesn’t show that all horses are animals. The issue is that it assumes at the beginning what it wants to show, and then concludes what we already knew. A proper line of reasoning would be the following
1. \(3 = 3\) is clear.
2. \(2 + 1 = 3\) is also clear.
3. Thus \(3y\) is the same as \((2 + 1)y\), since \(3\) and \(2 + 1\) are the same thing.
4. By distributing, this means \(3y\) and \(2y + y\) are the same.
5. Thus \(2y + y = 3y\).

Section I.2. Calculus Review

There are two fundamental rules for calculating derivatives: the chain rule, and the product rule. As a kind of inverse operation, these correspond to two rules of integration: \(u\)-substitution, and integration by parts respectively.

§ I.2.A. Differentiation

In most circumstances, differentiation relies on two rules and a few memorized derivatives.

<table>
<thead>
<tr>
<th>I.2.A • 1. Result 1 (the chain rule)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Let (f) and (g) be two differentiable functions, each with one input. Therefore the composition (h(x) := f(g(x))) has a derivative (h') given by (h'(x) = f'(g(x)) \cdot g'(x)).</td>
</tr>
</tbody>
</table>

There is a useful mnemonic to remember this result:

\[
\frac{d}{dx} f(g(x)) = \frac{df}{dg} \cdot \frac{dg}{dx},
\]

where the \(dgs\) “cancel out”. This mnemonic isn’t super helpful in practice, however, as it doesn’t tell you where these functions are evaluated. More properly, it should be written as in the chain rule.

Along with the chain rule is the product rule.

<table>
<thead>
<tr>
<th>I.2.A • 2. Result 2 (the product rule)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Let (f) and (g) be two differentiable functions, each with one input. Therefore the product (h(x) := f(x) \cdot g(x)) has a derivative (h') given by (h'(x) = f'(x) \cdot g(x) + f(x) \cdot g'(x)).</td>
</tr>
</tbody>
</table>

Both of these results should be very familiar to you, and they allow you to calculate the derivatives of complicated expressions just through syntactical manipulation. Often this happens without any actual thought required beyond remembering a few derivatives. In particular, the following will be important.

<table>
<thead>
<tr>
<th>I.2.A • 3. Result 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>You should know</td>
</tr>
</tbody>
</table>
| 1. \(
\frac{d}{dx} e^x = e^x\), i.e. \(e^x\) isn’t changed by differentiation; |
| 2. \(
\frac{d}{dx} x^n = nx^{n-1}\) for any fixed number \(n\) like 3, \(\pi\), \(\sqrt{1/2}\), etc.; |
| 3. \(
\frac{d}{dx} \ln |x| = 1/x;\) |
| 4. \(
\frac{d}{dx} \sin x = \cos x;\) and |
| 5. \(
\frac{d}{dx} \cos x = -\sin x.\) |
§I.2.B. Integration

Anyone who has taken calculus knows that Integration is much harder than differentiation. In practice, integration relies on two major rules, which are just the reverse of the differentiation rules: u-substitution, and integration by parts. The easier of the two methods is substitution.

I.2.B • 1. Result 4 (u-substitution)

Let \( f \) and \( u \) be smooth functions, each with one input. Therefore

\[
\int_a^b f(u(x)) \cdot u'(x) \, dx = \int_{u(a)}^{u(b)} f(u) \, du.
\]

This again helps show where exactly these things are evaluated, so that if we’re instead taking the indefinite integral, it’s not hard to see that \( \int f(u(x)) \cdot u'(x) \, dx \) is just \( F(u(x)) \) where \( F = \int f \, dx \).

The more difficult to master method of integration is integration by parts, a kind of inverse of the product rule. This is difficult partly because the form is less intuitive, but also because it requires more choice and calculation. Often one will need to do several attempts just to get the right choice of functions that allows the integral to be done easily. Occasionally, some further algebra is needed to get the value.

I.2.B • 2. Result 5 (integration by parts)

Let \( f \) and \( g \) be smooth functions, each with one input. Therefore

\[
\int_a^b f'(x) \cdot g(x) \, dx = f(x) \cdot g(x) \bigg|_a^b - \int_a^b f(x) \cdot g'(x) \, dx.
\]

As we can see above, there’s no change from where we’re evaluating the functions, so there is then no harm in stating the indefinite case as just

\[
\int f'g \, dx = fg - \int fg' \, dx.
\]

The difficulty then comes in choosing which function one wants to be \( f' \) as opposed to \( g \). This method requires that intuition, but also the ability to calculate the derivatives of these functions as well as their integrals, and then integrating their product.

Problems like these are the bread and butter of integration exercises. But often, integrals can be calculated in a slightly more direct way using complex numbers. Not all the time, of course, but in certain cases as we will see.

Section I.3. Complex Numbers Review

If the real numbers can be viewed as a line, the complex numbers can be viewed as a plane:
This means any complex number \( z \) is written as \( z = a + bi \), where \( a \) is referred to as the real part of \( z \), and \( b \) is the imaginary part of \( z \). In notation,

\[
z = \text{Re}(z) + i \cdot \text{Im}(z).
\]

And as with any other coordinate plane, we can represent any two points by cartesian coordinates, or by polar coordinates. For example, \(-1 + i\) is written in a kind of cartesian form: the sum of its real \((-1)\) and imaginary \((1)\) parts: \( p = (-1) + 1 \cdot i \). Alternatively, we can view any point \( p \) in \( \mathbb{C} \) by its angle with the (positive) real axis, and distance from 0.

Complex numbers can be brought about and defined in many ways: as a vector space, as an algebraic closure, as a quotient, and so on. The value \( i \) can be viewed as being anywhere from merely a formal symbol with no inherent connection to the real world to being a fundamental aspect in the algebra of even the real numbers. For our purposes, we have the following definitions.

### I.3.1. Definition 1

Extend the real numbers \( \mathbb{R} \) with a value \( i \) that satisfies \( i^2 = -1 \).

An **imaginary number** is a quantity \( ai \) for a real number \( a \).

A **complex number** is a quantity \( a + bi \) for real numbers \( a \) and \( b \). The set of complex numbers is \( \mathbb{C} \).

For \( z = a + bi \) a complex number,

- the real part of \( z \) is \( \text{Re}(z) = a \).
- The imaginary part of \( z \) is \( \text{Im}(z) = b \).
- The absolute value or magnitude of \( z \) is \( |z| = \sqrt{a^2 + b^2} \), a real number.
- The conjugate of \( z \) is \( \overline{z} = a - bi \).

The (multiplicative) inverse of a complex number \( z \) is a complex number \( s \) where \( s \cdot z = 1 \). If such an \( s \) exists, it’s called \( 1/z \).

Note that since \( s \cdot 0 = 0 \neq 1 \), this means 0 has no multiplicative inverse—i.e. we can’t divide by 0. Writing \( z = a + bi \) yields that

\[
\frac{1}{z} = \frac{1}{a + bi} \cdot \frac{a - bi}{a^2 - b^2i^2} = \frac{a - bi}{a^2 + b^2} = \frac{\overline{z}}{|z|^2}.
\]

Instead of the cartesian form \( a + bi \), we can also represent complex numbers in polar form: \( re^{i\theta} \). Proving this requires working with whatever definition of \( e \) one prefers. So when working with sines and cosines, Euler’s formula is a must-have.

### I.3.2. Result 6 (Euler’s formula)

For any \( \theta \), \( e^{i\theta} = \cos \theta + i \sin \theta \).

The usual exponentiation laws still hold: \( e^{x}e^{y} = e^{x+y} \). And so when \( n \) is an integer, \( (e^{x})^n = e^{x \cdot n} \).

But note that logarithms now require a choice, e.g. \( e^{\pi i} = -i = e^{3\pi i} \) so we must choose whether \( \ln i \) is \( \pi i \) or \( 3\pi i \) (or any of the other odd multiples of \( \pi i \)). Similarly, taking roots requires more choice than before, e.g. \( 1^4 = (-1)^4 = i^4 = (-i)^4 = 1 \), so we must be careful in taking the fourth root of 1, for example. As it turns out, such a choice is rarely important for our purposes, just as it is in the real case or in functions like \( \tan^{-1} \).

Returning to Euler’s formula, this implies \( |e^{i\theta}| = 1 \) for all \( \theta \). Furthermore, this gives a quick way to verify some trigonometric equalities, e.g.

\[
(e^{i\theta})^2 = (\cos \theta + i \sin \theta)^2 = \cos^2 \theta - \sin^2 \theta + i2 \sin \theta \cos \theta
\]
\[ e^{i2\theta} = \cos(2\theta) + i \sin(2\theta). \]

Equating the real and imaginary parts yields that \( \cos(2\theta) = \cos^2 \theta - \sin^2 \theta \), and \( \sin(2\theta) = 2 \sin \theta \cos \theta \).

The central question when a new concept is brought up like this is “what good is it?”. The answer for our purposes is mostly the ease of calculating integrals. This is accomplished by reducing the many hard-to-remember trigonometric identities to easy-to-confirm identities about \( e^z \) for complex \( z \), and by being able to factor any polynomial. For example, the double (and triple, half, etc.) angle formulas for \( \sin \) and \( \cos \) are easy to confirm using Euler’s formula. Following easily from Euler’s formula are the following important identities.

### I.3.3. Result 7

For every real number \( \theta \) and \( x \),

1. \[ \frac{d}{dx} e^{ix} = i \cdot e^{ix}; \]
2. \[ \cos \theta = \frac{1}{2} (e^{i\theta} + e^{-i\theta}) = \text{Re}(e^{i\theta}); \]
3. \[ \sin \theta = \frac{1}{2i} (e^{i\theta} - e^{-i\theta}) = \text{Im}(e^{i\theta}); \]

This means that we can easily turn powers of \( \cos \theta \) and \( \sin \theta \) into powers and sums of \( e^{i\theta} \) and \( e^{-i\theta} \), which can be far more easy to integrate, as we’ll see later. But beyond integration and \( e \), the use of \( \mathbb{C} \) is its ability to factor polynomials: any polynomial can be factored into factors of the form \( (x - c) \) with \( c \) complex.

In \( \mathbb{R} \), the polynomial \( x^2 + 4 \) cannot be factored. In \( \mathbb{C} \), however, it can be rewritten as \( x^2 - (2i)^2 \), which we know can be factored into \( (x + 2i)(x - 2i) \). The fact that any polynomial can be factored into terms of the form \( (x - c) \) is what makes the complex numbers so nice to work with in the context of algebra. As another example, we can find cube roots of 1 (that aren’t just 1). Using cartesian coordinates, this is difficult:

\[ z^3 = 1 \iff z^3 - 1 = 0 \iff (z - 1)(1 + z + z^2) = 0. \]

Factoring \( 1 + z + z^2 \) isn’t obvious. Using the polar idea is much easier:

\[ z^3 = 1 \iff (re^{i\theta})^3 = 1 \iff r^3 e^{i3\theta} = 1 \iff r^3 = 1 \text{ and } 3\theta = 2\pi k \text{ for some } k \in \mathbb{Z}. \]

Hence the solutions are \( z = e^{2\pi ik/3} \) for \( k \) in \( \mathbb{Z} \), which yields three solutions in total.
Section I.4. Worked-Out Examples

First, let’s calculate some integrals with traditional calculus methods.

I.4 • 1. Example 1

Calculate \( \int e^{3x} \, dx \).

Solution ..:
Here our choice of \( f' \) and \( g \) aren’t super clear, right? We need to fill out a table:

\[
\begin{align*}
  f' &= e^{3x} & f &= \frac{1}{3}e^{3x} \\
  g' &= 1 & g &= x
\end{align*}
\]

Using this, can calculate by integration by parts (I.2.B • 2)

\[
\int e^{3x} \, dx = \int f'g \, dx = fg - \int fg' \, dx
= \frac{1}{3}xe^{3x} - \int \frac{1}{3}e^{3x} \, dx = \frac{1}{3}xe^{3x} - \frac{1}{9}e^{3x}.
\]

I.4 • 2. Example 2

Calculate \( \int x^2 \sin x \, dx \).

Solution ..:
Here our choice of \( f' \) and \( g \) are a little less clear, in that it doesn’t seem so immediate what to choose. We again need to fill out a table:

\[
\begin{align*}
  f' &= \sin x & f &= -\cos x \\
  g' &= 2x & g &= x^2
\end{align*}
\]

Using this, can calculate by integration by parts (I.2.B • 2)

\[
\int x^2 \sin x \, dx = \int f'g \, dx = fg - \int fg' \, dx
= (\cos x)x^2 - \int -2x \cos x \, dx.
\]

At this point, it’s not entirely clear how to integrate \( \int x \cos x \, dx \). So we will need to use integration by parts (I.2.B • 2) again: first we fill out a table:

\[
\begin{align*}
  u' &= \cos x & u &= \sin x \\
  v' &= 1 & v &= x
\end{align*}
\]

And so we can calculate \( \int x \cos x \, dx \) (and from there \( \int -2x \cos x \, dx \) to yield \( \int x^2 \sin x \, dx \)) through

\[
\int x \cos x \, dx = \int u'v \, dx = uv - \int uv' \, dx
= x \sin x - \int \sin x = x \sin x + \cos x.
\]

Using this in the above calculation, we get

\[
\int x^2 \sin x \, dx = (\cos x)x^2 - \int -2x \cos x \, dx
= -x^2 \cos x + 2(x \sin x + \cos x)
= (2 - x^2) \cos x + 2x \sin x.
\]
Using complex numbers and functions is very natural. Integrals that traditionally might have required remembering a trigonometric identity and using it cleverly can be made much more straightforward through the use of complex functions and Euler’s formula (I.3 • 2). For example, consider the integral of $\cos^2 x$. Normally, you would need to use the trigonometric identity $\sin^2 x = 1 - \cos^2 x$. But if this was forgotten, a reasonable approach would be the following.

### I.4 • 3. Example 3

**Calculate $\int \cos^2 x \, dx$.**

**Solution:**

Here our choice of $f'$ and $g$ is pretty clear: both should be $\cos x$. But then we need to fill out a table:

- $f' = \cos x$  
  $f = \sin x$
- $g' = -\sin x$  
  $g = \cos x$

Hence we get

$$\int \cos^2 x \, dx = \int f'g \, dx$$

$$= fg - \int fg' \, dx$$

$$= \sin x \cos x - \int \sin x(-\sin x) \, dx$$

$$= \sin x \cos x + \int \sin^2 x \, dx.$$

Now we want to find the integral of $\int \sin^2 x \, dx$. So we again must do integration by parts, and fill out a table:

- $h' = \sin x$  
  $h = -\cos x$
- $k' = \cos x$  
  $k = \sin x$

Hence we get

$$\int \sin^2 x \, dx = \int h'k \, dx = hk - \int hk' \, dx$$

$$= -\cos x \sin x - \int (-\cos x) \cos x \, dx$$

$$= -\sin x \cos x + \int \cos^2 x \, dx.$$

Plugging this back into our formula for $\int \cos^2 x \, dx$ gives us nothing new:

$$\int \cos^2 x \, dx = \sin x \cos x - \sin x \cos x + \int \cos^2 x \, dx = \int \cos^2 x \, dx.$$

We can see that this approach fails. But where this method failed, using Euler’s formula (I.3 • 2) works.

### I.4 • 4. Example 4

**Calculate $\int \cos^2 x \, dx$.**

**Solution:**

Recall that $\cos x = \frac{1}{2}(e^{ix} + e^{-ix})$. Hence when we square, we get the sum of easily integrable functions:

$$\int \cos^2 x \, dx = \int \frac{1}{4}(e^{ix} + e^{-ix})^2 \, dx$$

$$= \int \frac{1}{4}(e^{2ix} + 2 + e^{-2ix}) \, dx$$
In simplifying this, we can either use Euler’s formula (I.3 • 2), or recognize \( \sin 2x \) inside there. Either way gives the same result:

\[
\int \cos^2 x \, dx = \frac{1}{4} \left( \sin(2x) + 2x \right) + c.
\]

Of course, if one did remember the trigonometric identity, this way seems a little more involved. But a simple change from \( \cos^2 x \) to \( \cos^3 x \), or higher, makes the problem even more difficult when going the trigonometric route. Even if computationally tedious, using Euler’s formula (I.3 • 2) is much more straightforward. We can consider another example comparing the two methods.

I.4 • 5. Example 5

Calculate \( \int e^x \sin x \, dx \).

Solution ..:

Unlike the previous example, we don’t need to remember many trigonometric identities to use the traditional method. We do, however, need to have some creativity and intuition about what to do. Using integration by parts, we will fill out the table

\[
\begin{array}{c|c}
\text{\( f' \)} & \text{\( f \)} \\
\hline
\sin x & -\cos x \\
\hline
\end{array}
\]

\[
\begin{array}{c|c}
\text{\( g' \)} & \text{\( g \)} \\
\hline
e^x & e^x \\
\hline
\end{array}
\]

Hence we have the equality

\[
\int e^x \sin x \, dx = \int f'g \, dx = fg - \int fg' \, dx = -e^x \cos x + \int e^x \cos x \, dx.
\]

Now we need to calculate \( \int e^x \cos x \, dx \). Again, the only real option available to us is integration by parts. Now we must again make some decisions on how to proceed, the second place intuition is required.

\[
\begin{array}{c|c}
\text{\( u' \)} & \text{\( u \)} \\
\hline
\cos x & \sin x \\
\hline
\end{array}
\]

\[
\begin{array}{c|c}
\text{\( v' \)} & \text{\( v \)} \\
\hline
e^x & e^x \\
\hline
\end{array}
\]

Hence we have the equality

\[
\int e^x \cos x \, dx = \int u'v \, dx = uv - \int uv' \, dx = e^x \sin x - \int e^x \sin x \, dx.
\]

At this point, it may appear that we’re stuck: in trying to calculate \( \int e^x \sin x \, dx \), it seems like we need to have already calculated it. In the third occurrence of intuition, we can realize the equality our first use of integration by parts gave us:

\[
\int e^x \sin x \, dx = -e^x \cos x + \int e^x \cos x \, dx
\]

\[
= -e^x \cos x + e^x \sin x - \int e^x \sin x \, dx
\]

\[
\therefore 2 \int e^x \sin x \, dx = e^x \cos x + e^x \sin x + c
\]

\[
\int e^x \sin x \, dx = \frac{1}{2} e^x \cos x + \frac{1}{2} e^x \sin x + c.
\]

The \( +c \) comes from the fact that we’re using indefinite integrals: the \( \int e^x \sin x \, dx \) on the left may differ from \( \int e^x \sin x \, dx \) on the right by a constant. But the important thing is that we found our answer through a long process.

A method which is much more mechanical, requiring fewer choices, and less intuition, is the following method:
Calculate $\int e^x \sin x \, dx$.

**Solution:**

We can use either the identity of $\sin x = \text{Im}(e^{ix})$, or $\sin x = \frac{1}{2i} \left( e^{i\theta} - e^{-i\theta} \right)$. Either will work, but to showcase a different method, we’ll consider the identity $\sin x = \text{Im}(e^{ix})$, as witnessed by Euler’s formula (I.3 • 2). The is the only part of the calculation which requires some intuition. But what it means is that

$$\int e^x \sin x \, dx = \int e^x \text{Im}(e^{ix}) \, dx = \int \text{Im}(e^z e^{iz}) \, dz = \text{Im} \left( \int e^z e^{iz} \, dz \right)$$

So the work of calculating the integral is already done. The only work left is just expanding definitions and so forth to find the imaginary part. This means that we need to put the inside part $\frac{1}{1+i} e^{(1+i)z}$ into that cartesian form $a + bi$. Through the usual methods of expanding, using Euler’s formula (I.3 • 2), and getting square roots out of the denominator,

$$\frac{1}{1+i} e^{(1+i)z} = \frac{1}{1+i} \cdot \frac{1-i}{1-i} e^z e^{iz}$$

$$= \frac{1-i}{1-i^2} e^z e^{iz}$$

$$= e^z \frac{1-i}{2} (\cos z + i \sin z)$$

$$= \frac{1}{2} e^z (\cos z + i \sin z - i \cos z + \sin z)$$

$$= \frac{1}{2} e^z (\cos z + \sin z + i \frac{1}{2} e^z (\sin z - \cos z).$$

This is of the form $a + bi$, and so its imaginary part is $\frac{1}{2} e^z (\sin z - \cos z)$, which means

$$\int e^x \sin x \, dx = \frac{1}{2} e^z (\sin z - \cos z) + c.$$
solutions are $\sqrt[3]{3}e^{i(1+2k)\pi/4}$ for $k$ as 0, 1, 2, and 3.

Returning to the original question, the factorization is then

\[ x^4 + 3 = \left(x - \sqrt[3]{3}e^{i\pi/4}\right)
\left(x - \sqrt[3]{3}e^{i3\pi/4}\right)
\left(x - \sqrt[3]{3}e^{i5\pi/4}\right)
\left(x - \sqrt[3]{3}e^{i7\pi/4}\right). \]

**Example 8**

Use Euler’s formula to write $e^{2-3i}$ in the form $a + bi$ for $a$ and $b$ real numbers.

**Solution**:

\[ e^{2-3i} = e^2e^{-3i} = e^2(\cos(-3) + i\sin(-3)) = e^2 \cos 3 - ie^2 \sin 3. \]

**Example 9**

Use Euler’s formula to write $2^{1-i}$ in the form $a + bi$ for $a$ and $b$ real numbers.

**Solution**:

\[ 2^{1-i} = 2 \cdot 2^{-i} = 2e^{-i\ln(2)} = 2(\cos(-\ln 2) + i\sin(-\ln 2)) = 2\cos(\ln 2) - i \cdot 2\sin(\ln 2). \]
Chapter II. First-Order, Differential Equations

Section II.1. Basics

Unknowingly, you’ve come across differential equations before in calculus. One of the simplest differential equations would be $y' = 0$, which implies that $y$ is just a constant. Similarly, if $y' = a$ for some constant $a$, we get that $y(t) = at + c$ for some constant $c$. Thus the solution to $y' = a$ is not unique: for different $c$, we get different $y = at + c$ satisfying the differential equation. In general, we’ve seen the differential equation $y' = f(t)$, in which case the general solution is the (indefinite) integral of $f$.

To generalize this, we will consider what happens when $y$ itself is related to its derivative. This means differential equations of the form $y' = f(t, y)$ for some function $f$ of $t$ and $y$. Note that there might not be any solutions to an arbitrary differential equation. For example, $y' = y + 1$ has no solution for $y$. Equations are statements about functions, and so are sometimes true and sometimes false. Our goal is to find out when they are true. This can be done by examining the equations themselves. Again, let’s consider equations of the form $y' = f(t, y)$ for some function $f$. Using this, we can see what $y'$ would need to be if a solution $y$ had a value $y_0$ at $t_0$. Plotting this gives a direction field.

§ II.1.A. Direction fields

Direction fields are primarily a visual aid in that they give a general idea how any given solution to a differential equation behaves. You can think of these fields as acting like currents of a river: if you plop in at some initial value, the path the direction field carries you is the solution with that initial value. These fields are easily displayed when we have $y'$ equal to an expression involving just $y$ and $t$:

$$y' = f(t, y),$$

for some function $f$. The idea is that each point $(a, b)$ in the $ty$-plane then has a corresponding direction it goes according to the above equation. In practice, this means drawing a small line with slope $f(a, b)$ at $(a, b)$.

Again, this is mostly a visual aid, allowing you to see at a glance whether solutions will tend towards a certain value or not just by starting from a variety of initial values, and following the lines of the diagram. For example, consider the differential equation $y' = y^2 + \frac{1}{2}y - \frac{1}{2}t - 1$. If we consider the values of $y'$ when $y$ and $t$ are between $-3$ and $3$, we can get a variety of slopes. These slopes are then used as the slopes of arrows, giving the direction field below. Two solutions are also plotted to show what the solutions look like.

---

1 In particular, this $c$ is just $y(0) = a \cdot 0 + c = c$.

2 Later results tell us that there is in fact only one such solution (when $f$ is continuously differentiable with respect to $y$) with that initial value, so the use of “the” here is appropriate.
There are many different solutions to this differential equation, like most, since any given solution might have different initial values. For example, \( y' = 1 \) has solutions \( t \), but also \( t + 2 \), and \( t + c \) for any \( c \). Remembering constants of integration will turn out to be very important here. Whereas in previous calculus courses such a constant wasn’t very important, here it can be crucial.

### § II.1.B. Descriptions of differential equations

Sometimes it’s easy to solve certain kinds of differential equations. As such, it’s useful to have names to reference such equations. Additionally, working with a smaller subset of differential equations can be easier than dealing with them in general. The same holds for us. For our purposes, we have the following concepts and descriptions.

<table>
<thead>
<tr>
<th>II.1.B • 1. Definition 2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A differential equation</td>
</tr>
<tr>
<td>1. is ordinary if it doesn’t involve partial derivatives.</td>
</tr>
<tr>
<td>2. is partial if it involves partial derivatives.</td>
</tr>
<tr>
<td>3. is ( n )th order if it involves the ( n )th derivative (without any higher derivatives).</td>
</tr>
</tbody>
</table>

The first kinds of differential equations we will be looking at will be separable, and linear.

<table>
<thead>
<tr>
<th>II.1.B • 2. Definition 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>A differential equations is separable if it is of the form (or at least can be put in the form)</td>
</tr>
<tr>
<td>( y' = f(t)g(y) ),</td>
</tr>
<tr>
<td>for some functions ( f ) and ( g ).</td>
</tr>
</tbody>
</table>

A general process for solving separable equations is then to divide by \( g(y) \), yielding \( y' / g(y) = f(t) \). Integrating then gives an equations involving just \( y \) and \( t \), with no derivatives.

<table>
<thead>
<tr>
<th>II.1.B • 3. Definition 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>A differential equation is linear if it is of the form (or at least can be put in the form)</td>
</tr>
<tr>
<td>( a_0(t)y + a_1(t) \frac{dy}{dx} + \cdots + a_n(t) \frac{d^n y}{dx^n} = g(t) ),</td>
</tr>
<tr>
<td>for some functions ( g, a_0, \ldots, a_n ).</td>
</tr>
</tbody>
</table>

One simple test to see whether an equation is linear to see whether there are any terms like \( y' \cdot y \) or \( y''' \cdot y' \). If there are—and you can’t just divide by or subtract them to remove them from both sides—then the equation is not linear. Over time, we will look at more types of differential equations.

### § II.1.C. Initial value problems

As above, there can be many solutions to a differential equation. Specifying an initial value just means specifying the solution. Usually, you can end up with a general form for the solution involving some unknowns like a + c from an integral: \( y(t) = f(t, c) \), for some function \( f \). If this is the general solution, specifying the initial value \( y(0) = y_0 \) requires \( y_0 = f(0, c) \), which often is enough to uniquely identify \( c \). Note that 0 isn’t particularly special, as we could just as easily specify the value at 1, \( y(1) = y_1 \), and so require \( c \) to satisfy \( y_1 = f(1, c) \).
Either way, solving for \( c \)—and plugging this back into the general form—gives us an explicit way to calculate \( y \) for any value of \( t \).

§ II.1.D. Two simple initial value problems

One of the simplest examples of interaction between \( y' \) and \( y \) is just equality.

\[
y' = y.
\]

This equation is both separable, and linear, and so the solutions of this can be found just by dividing by \( y \) and then integrating. If \( y \neq 0 \) somewhere, then

\[
y' = y \quad \rightarrow \quad \int \frac{1}{y} \, dy = 1 \quad \rightarrow \quad \ln |y| + C_1 = t + C_2
\]

\[
\ln |y| = t + C_3 \quad \rightarrow |y| = e^{t+C_3} = e^{C_3} e^t.
\]

By choosing a new constant\(^{iii}\)

\[
C = \begin{cases} 
  e^{C_3} & \text{if } y > 0 \text{ somewhere} \\
  -e^{C_3} & \text{if } y < 0 \text{ somewhere} \\
  0 & \text{if } y = 0 \text{ somewhere}.
\end{cases}
\]

we can get rid of the absolute value around \( y \) to get the equality

\[
y = Ce^t,
\]

for some constant \( C \). In fact, note that \( y(0) = Ce^0 = C \), so that we can actually say \( y = y(0)e^t \). Such a derivation is common with simple equations like this, and it is one of the reasons we will encounter \( e \) so much.

Let’s consider a slightly more complicated relationship between \( y' \) and \( y \):

\[
y' = ay - b.
\]

For various \textit{numbers} \( a \) and \( b \). Later we’ll encounter more complicated expressions where \( y' = f(t)y - g(t) \) for functions \( f \) and \( g \). For now, this simpler equation can be solved in the same way each time, in a similar way as before.

II.1.D • 1. Result 8

Let \( a \neq 0 \). Let \( y \) be a function satisfying \( y' = ay - b \), with \( y(t_0) = w \), some specified value. Therefore for any \( t \),

\[
y(t) = \left( w - \frac{b}{a} \right) e^{a(t-t_0)} + \frac{b}{a}.
\]

\[\text{Proof . . .} \]

Note that \( y' = ay - b \) implies \( y' = a(y - b/a) \), and hence

\[
\frac{y'}{y - b/a} = a.
\]

If we integrate both sides with respect to \( t \), we get

\[
\int \frac{y'(t)}{y(t) - b/a} \, dt = at + c
\]

for some constant \( c \). Through \( u \)-substitution, we can evaluate the integral on the left. So for all \( t \),

\[
\ln |y(t) - b/a| = at + c.
\]

\(^{iii}\)This is a constant, since if \( y > 0 \) somewhere, then \( y > 0 \) everywhere, and similarly for \( y < 0 \). To see this, since \( y \) is continuous, otherwise \( y \) would need to go from being positive to negative, and so be \( 0 \) somewhere. But for \( y \) to be \( 0 \) somewhere requires \( y \) to be \( 0 \) everywhere by some later results about the uniqueness of solutions.
We can then solve for \( y(t) \) to get that
\[
|y(t) - \frac{b}{a}| = e^{t}e^{-a} \quad \longrightarrow \quad y(t) = Ce^{at} + \frac{b}{a}.
\]
This \( C \) is just some constant—in particular, it’s short-hand for \( e^{c} \cdot \text{sign}(w - \frac{b}{a}) \) given the \( c \) above. Calculating this \( C \) can be done with the initial value \( w \):
\[
w = y(t_0) = Ce^{at_0} + \frac{b}{a} \quad \longrightarrow \quad w - \frac{b}{a} = Ce^{at_0}.
\]
Hence, we can write
\[
y(t) = \left( w - \frac{b}{a} \right) e^{a(t-t_0)} + \frac{b}{a},
\]
which of course, presupposes \( a \neq 0 \). If \( a = 0 \), then the solution is trivial: \( y' = b \) implies \( y \) is just a line with slope \( b \) through the point \( (t_0, w) \).

In special cases, this complicated expression for \( y \) can be simplified quite a bit, e.g. when \( t_0 \) or \( w = 0 \).

Problems involving these differential equations come in a few forms. The first is to go through the process above in a particular case. Another is to find a value of \( t \) where \( y(t) = m \) for some value \( m \). This can be done by finding the solution as above, and then solving for \( t \) in the equation \( y(t) = m \).

§ II.1.E. Existence and uniqueness theorems

The existence and uniqueness theorems are important for actual mathematics, but here we’re focused on doing calculations rather than doing math. So for our purposes, these theorems merely confirm that we don’t need to do extra work in finding solutions. Once we’ve found a solution that works, that’s the only solution: there’s no need to look any further to see if maybe we’re missing something. Of course, the theorems don’t apply in all cases, so we must be careful in general.

Of the several results the book[1] gives, the existence and uniqueness of a solution to a differential equation is satisfied so long as the given functions are continuous—or in some sense smooth—in a given area, which guarantees the existence and uniqueness for at least a part of that area.

---

### II.1.E.1. Result 9 (The Existence and Uniqueness Theorem)

Let \( f \) and \( \partial f / \partial y \) be continuous in some rectangle \( R \). Therefore there is exactly one \( y \) such that \( y(t_0) = y_0 \) and
\[
y' = f(y, t),
\]
where \( y \) is defined on an interval around \( t_0 \) within the bounds of \( R \).

This means that we have existence and uniqueness locally, although perhaps not globally. As an analogy, the equation for (the graph of) the unit circle is \( x^2 + y^2 = 1 \). This does not define a function \( y \), since there will be two \( y \)s for any \( x \). But if we restrict our view to \( x \) in the interval \([-1, 1]\) and \( y \) in the interval \([0, 1]\), it does define a function: \( y = \sqrt{1 - x^2} \) over \((-1, 1)\).

Section II.2. Linear Equations

In the previous section, we worked with differential equations of the form
\[
y' = ay + b,
\]
for constants \( a \neq 0 \) and \( b \). This can be generalized slightly, where \( a \) and \( b \) are instead functions: \( y' = f(t)y + g(t) \).

\[iv\] if \( a = 0 \), the differential equation is just \( y' = b \), which is easily solved: \( y(t) = bt + y(0) \).
In a slightly more usable form, we can write how this will usually be written:

\[ y' + p(t)y = g(t) \]

for functions \( p \) and \( g \). Unlike the simpler equation, this is not separable, so our method to solve the equation cannot be so simple as just (dividing and then) integrating both sides. Instead we have two methods of finding solutions.

### § II.2.A. Integrating factors

The method of integrating factors is just finding a \( \mu \neq 0 \) that satisfies

\[ (\mu(t)y)' = \mu(t)y' + \mu'(t)y, \quad \text{i.e. that satisfies} \quad \mu'(t) = \mu(t)p(t). \]

This method allows us to conclude the following.

#### II.2.A.1. Result 10

Let \( p \) and \( g \) be continuous functions, and consider the differential equation \( y' + p(t)y = g(t) \) with \( y(t_0) = y_0 \). Therefore

\[ \mu(t) = e^{\int_{t_0}^{t} p(x) \, dx} \]

is an integrating factor, and

\[ y(t) = \frac{y_0}{\mu(t)} + \frac{1}{\mu(t)} \int_{t_0}^{t} \mu(x)g(x) \, dx. \]

**Proof.**

The key observation is that \( y' + p(t)y \) looks a bit like the product rule:

\[ (\mu(t)y)' = \mu(t)y' + \mu'(t)y. \]

If we can find a \( \mu \) where \( \mu' = \mu p(t) \), then the equation becomes a simple matter of integrating both sides again, and then dividing by \( \mu(t) \). To solve \( \mu' = \mu p(t) \), note that then

\[ \frac{\mu'}{\mu} = p(t). \]

Hence the integrals are equal: \( \int_{t_0}^{t} \frac{\mu'}{\mu} \, dt = \int_{t_0}^{t} p(t) \, dt \). As a result, a \( u \)-substitution of \( \mu(t) \) yields

\[ \ln |\mu(t)| = \int_{t_0}^{t} p(t) \, dt + C \]

for some constant \( c \). Note that then

\[ \mu(t) = e^{\int_{t_0}^{t} p(t) \, dt} \cdot \mathcal{C}. \]

is then the solution, for some value of \( \mathcal{C} \). It’s important to realize now that we were just trying to find some function \( \mu \) that would satisfy the equation. So any choice of \( \mathcal{C} \), excluding 0, works. To make things simpler, we will choose \( \mathcal{C} = 1 \), making \( \mu(t_0) = C = 1 \). As a result \( \mu(t) \) is always positive, and so we can divide and multiply by it:

\[ g(t) = y' + p(t)y \quad \leftrightarrow \quad \mu(t)g(t) = \mu(t)y' + \mu(t)p(t)y \]

\[ \leftrightarrow \quad \mu(t)g(t) = \mu(t)y' + \mu'(t)y \]

\[ \leftrightarrow \quad \mu(t)g(t) = (\mu(t)y)' \]

\[ \leftrightarrow \quad \int_{t_0}^{t} \mu(t)g(t) \, dt = \mu(t)y(t) - \mu(t_0)y_0 \]

\[ \leftrightarrow \quad \int_{t_0}^{t} \mu(t)g(t) \, dt = \mu(t)y(t) - y_0 \]

\[ \leftrightarrow \quad y_0 + \int_{t_0}^{t} \mu(t)g(t) \, dt = \mu(t)y(t) \]
that way.

As a result, the method of integrating factors allows one to solve all first-order, linear, ordinary differential equations.

§ II.2.B. First-order undetermined coefficients

Integrating factors can be a slow process, however, requiring us to integrate potentially very complicated functions. A slightly easier idea is contained in a kind of guess and check. This method is further elaborated on in Subsection IV.2.B. For now, consider the version of the differential equation where \( g(t) = 0 \), and \( p(t) = a \) is some constant.

\[
y' + ay = 0.
\]

Note that any two solutions to \( y' + ay = g(t) \) will have their difference satisfy \( y' + ay = 0 \). As a result, if \( \phi(t) \) is the general solution to \( y' + ay = 0 \), and \( Y(t) \) is any solution to \( y' + ay = g(t) \), then we can say that the general solution to \( y' + ay = g(t) \) will be \( y(t) = \phi(t) + Y(t) \). Solving \( y' + ay = 0 \) is relatively easy, since it’s separable: \( y' = -ay \) and hence the general solution is \( y(t) = Ce^{-at} \).

To find a particular solution \( Y(t) \), we can do guess-and-check based on \( g(t) \). This is more restrictive than integrating factors, but can sometimes be much easier to do in practice.

<table>
<thead>
<tr>
<th>II.2.B • 1. Result 11 (Undetermined Coefficients)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Consider the differential equation ( ay' + by = g(t) ).</td>
</tr>
<tr>
<td>If ( g(t) = at^n + \cdots + a_1 t + a_0 ), then guess ( Y(t) = A_n t^n + \cdots + A_1 t + A_0 );</td>
</tr>
<tr>
<td>If ( g(t) = \sin(at) ), or ( g(t) = \cos(at) ), then guess ( Y(t) = A \sin(at) + B \cos(at) );</td>
</tr>
<tr>
<td>If ( g(t) = e^{at} ), then guess ( Y(t) = Ae^{at} );</td>
</tr>
<tr>
<td>for various constants.</td>
</tr>
<tr>
<td>If the guess ends up being a solution to ( ay' + by = 0 ), then just multiply the guess by ( t ). And if that also is a solution, keep multiplying by ( t ) until it isn’t.</td>
</tr>
</tbody>
</table>

If \( g \) is a product/sum of these, then our guess should be a product/sum of the corresponding guesses. For example, if \( g(t) = t \cos(2t) + 5e^t \), we should guess

\[
Y(t) = (At + B)(C \cos(2t) + D \sin(2t)) + Ee^t,
\]

for undetermined coefficients \( A, B, C, D, \) and \( E \). Note we need to use different constants for each guess involved.

Once we have our guess, it’s a matter of figuring out what constants work, which means plugging in \( Y(t) \) into \( ay' + by = g(t) \), and solving for the coefficients of \( Y \).

Because this is just one equation with many unknowns, it’s not obvious how we can solve for many different coefficients. To get around this, note that such an equation is supposed to hold for all \( t \), and thus we can plug in various values of \( t \) to get as many equations as we need. Alternatively—and more easily—we can equate coefficients, and get equations that way.

Section II.3. Separable and Autonomous Equations

\(^{\text{So long as the functions involved are sufficiently nice, which they will be in almost all practical applications.}}\)
§II.3.A. Separable equations

There’s another way to generalize the differential equations worked with in Section II.1:

\[ y' = h(y) \cdot g(t). \]

In the case of \( y' = ay + b \), \( g(t) \) is just a constant. By dividing by \( h(y) \), when this is non-0, a more workable form for the equation is

\[ f(y)y' = g(t). \]

Such equations are called separable, since we have, in some sense, separated the variables \( y \) and \( t \). Solving the differential equation is then just done by integration: using a \( u \)-substitution of \( y \):

\[ \int f(y) \, dy = \int g(t) \, dt. \]

In practice, we have the following result.

---

II.3.A • 1. Result 12

Let \( f \) and \( g \) be continuous functions such that

1. There is a function \( y \) satisfying \( f(y)y' = g(t) \); and
2. \( y(t_0) = y_0 \).

Therefore \( F(y(t)) - F(y_0) = \int_{t_0}^{t} g(t) \, dt \).

Proof: :

By integrating both sides, we get through a \( u \)-substitution of \( y \),

\[ f(y)y' = g(t) \iff \int_{y_0}^{y(t)} f(y) \, dy = \int_{t_0}^{t} g(t) \, dt \]

\[ F(y(t)) - F(y_0) = \int_{t_0}^{t} g(t) \, dt \]

Because the method is so simple, separable equations are useful to have. Unfortunately, the integration of the functions can be difficult. Similarly, taking the inverse of \( F \) to get a unique \( y \) might also be difficult to achieve, or possibly impossible. Often it can be useful to think of \( y \) not as a function, but as another variable in relation to \( t \), in particular one of potentially many \( y \) satisfying

\[ F(y) = F(y_0) + \int_{t_0}^{t} g(t) \, dt. \]

The resulting graph is then a path through the plane, but it might not be the graph of a function—i.e. it fails the vertical line test.

§II.3.B. Autonomous equations

To make things a little simpler, we can consider differential equations where there is no interaction from \( t \).

---

II.3.B • 1. Definition 5

A differential equation is autonomous if it is of the form

\[ y' = f(y) \]

for some function \( f \).

This means autonomous equations are separable, and so we can apply the methods there to solve the differential equations. Much of the book\[1\], however, is mostly focused on the stability of various constant solutions. In essence, if we find a constant \( c \) such that \( f(c) = 0 \), then the constant function \( y(t) = c \) satisfies the differential equation above:
$y'(t) = 0$, and $f(y(t)) = f(c) = 0$ so that $y' = f(y)$. This doesn’t necessarily hold in general, since in other cases $y'$ might depend on $t$. But for autonomous equations, $y'$ only depends on $y$.

### II.3.B • 2. Definition 6

Consider the differential equation $y' = f(y)$ for some $f$. The constant solution, critical point of $f$, or equilibrium, $y = c$ is

1. stable iff solutions with initial values “near” $c$ approach $c$ as $t \to \infty$,
2. unstable iff solutions with initial values “near” $c$ diverge from $c$ as $t \to \infty$,
3. semistable iff solutions near $c$ converge on one side, and diverge on the other.

Being “near” $c$ is a pretty ambiguous statement, but in most cases it’s clear what’s meant. For instance, if we have the equation

$$y'(t) = (y - 1)^2(y - 2)(y - 3),$$

then we have three constant solutions: $y(t) = 1$, $y(t) = 2$, and $y(t) = 3$. Being “near” the solution $y(t) = 2$ in this case just means being somewhere between 1 and 2, or 2 and 3. But being “near” $y(t) = 1$ means just being anywhere from $-\infty$ to 1, or from 1 to 2.

Finding out whether a given constant solution is stable, unstable, or semistable is often done just by looking at whether $y'$ is positive or negative near that value. Continuing the previous example,

$$y'(t) = (y - 1)^2(y - 2)(y - 3)$$

is positive for $y < 1$. Hence solutions near 1 converge to 1 when they are below it. Solutions between 1 and 2 yield $y'$ as positive. Thus solutions near 1 diverge from 1 when they are above it. Since they converge from below, and diverge when above, $y = 1$ is a semistable equilibrium.

### Section II.4. Exact Equations

Exact equations are a kind of generalization of separable equations, but where separable meant something of the form

$$f(y)y' + g(t) = 0,$$

exact equations allow $f$ and $g$ to be functions of both $y$ and $t$:

$$f(y, t)y' + g(y, t) = 0,$$

where $f(y, t)$ and $g(y, t)$ are partial derivatives of some other given function.

### II.4 • 1. Definition 7

A differential equation is exact iff it can be put in the form

$$\frac{d}{dt}F(y, t) = \frac{\partial F}{\partial y} \frac{dy}{dt} + \frac{\partial F}{\partial t} = 0,$$

for some $F(y, t)$.

From the multi-variable chain rule, this just states that

$$\frac{d}{dt}F(y, t) = 0,$$

i.e. $F(y, t) = c$,

for some constant $c$. Solving the differential equation is then the same as solving $F(y, t) = c$ for $y$.

Just like with separable equations, it might not be possible to get a unique $y$ from this. Nevertheless, such a form can allow us to describe solutions. More important for our purposes is how to find such an $F$. Supposing that we’re given nice enough functions, we can apply the following result which tells us when an equation is exact, and how to find the function in question. Note that we’re denoting partial derivatives by subscripts here: $f_i = \partial f / \partial t$ for example.
II.4.2. Result 13

Let \( f \) and \( g \) be continuously differentiable functions. Therefore
\[
f(y, t)y' + g(y, t) = 0 \text{ is exact } \iff f_t = g_y.
\]

Figuring out whether the equation is exact can be done fairly easily thanks to this pretty simple test: the equation is exact iff \( f_y = g_t \), referring to partial derivatives again. Once we know that an equation is exact, finding such an \( F \) is a problem of integration. In essence, we have
\[
F = \int f(y, t) \, dt + c(y).
\]

So now we regard the (no-longer) constant \( c \) from integration as a function of \( y \). To find this \( c(y) \), we integrate \( g(y, t) \):
\[
c(y) = \int g(y, t) \, dt - \int f(y, t) \, dt.
\]

There’s nothing special about \( f \) versus \( g \) here. We could just as easily do the reverse order, but the point is the equality
\[
F(y, t) = \int f(y, t) \, dt + c_1(y) = \int g(y, t) \, dy + c_2(t),
\]
for functions \( c_1 \) and \( c_2 \). This also makes intuitive sense since taking the partial derivatives makes these \( c_1 \) and \( c_2 \)s disappear, leaving just \( f(y, t) \) or \( g(y, t) \). Note again that such an \( F \) is not unique, since \( F + 2 \) or \( F + 15 \) works just as well as a witness to the fact that \( f + g y' = 0 \) is an exact differential equation. But we’re not interested in finding a unique \( F \), since just one will suffice to solve the differential equation\(^{vi}\).

The difficulty with exact equations is that they are really rare. But once you know that the equation is exact, then the differential equation can be dealt with fairly easily\(^{vii}\). Now sometimes we can turn a non-exact differential equation into an exact differential equation through the method of integrating factors: multiplying by some non-0 \( \mu(t) \)—solving another differential equation to find such a \( \mu \)—and then solving the new exact equation. The idea is similar to the integrating factors used for linear equations, except now more general.

II.4.3. Result 14

Consider the non-exact differential equation
\[
M(t, y)y' + N(t, y) = 0.
\]

An integrating factor \( \mu \) that turns this into an exact differential equation is one where
\[
(\mu M)y' + (\mu N) = 0,
\]
and which satisfies the partial differential equation
\[
\mu_t M + \mu M_t - \mu_y N - \mu N_y = 0.
\]

This result allows us to conclude that if \( \mu \) is a function of just \( t \), then
\[
\mu_t = \mu' = \frac{N_y - M_t}{M},
\]
allowing us to solve the differential equation to find an integrating factor \( \mu \). Similarly, if \( \mu \) is a function of just \( y \), then
\[
\mu_y = \mu' = \frac{M_t - N_y}{N}.
\]

Often these can be useful for confirming that we can find such a \( \mu \) by instead seeing whether something like
\[
\frac{M_t - N_y}{N}
\]
is a function of just \( t \) or just \( y \). But most of the time, finding such an integrating factor is not at all easy, and the above ideas won’t always work.

\(^{vi}\) and the difference of a constant won't matter, since this constant could be subtracted from the equation \( F(y, t) = c \) to yield merely a different arbitrary constant on the right hand side.

\(^{vii}\) at least conceptually
Section II.5. Worked-Out Examples

II.5.1. Example 10

Consider the differential equation $y'' + 2y = e^{-2t}$ with initial condition $y(0) = 2$. Solve for $y(t)$ using undetermined coefficients.

Solution .:

The solution $y(t)$ will be the sum $\phi(t) + Y(t)$ where $\phi' + 2\phi = 0$, and $Y' + 2Y = e^{-2t}$. First let’s find $Y(t)$. As in Undetermined Coefficients (II.2.B • 1), guess

$$Y(t) = Ae^{-2t}$$

$$Y'(t) = -2Ae^{-2t}.$$ 

Hence $Y(t)$ is a particular solution iff

$$e^{-2t} = Y'(t) + 2Y(t) = -2Ae^{-2t} + 2(Ae^{-2t}) = 0.$$ 

So our guess was accidentally a solution to $y' + 2y = 0$. So we should multiply our guess by $t$:

$$Y(t) = Ae^{-2t}t$$

$$Y'(t) = -2Ate^{-2t} + Ae^{-2t}.$$ 

Now we have that this is a particular solution iff

$$e^{-2t} = Y'(t) + 2Y(t) = -2Ate^{-2t} + Ae^{-2t} + 2Ate^{-2t}$$

$$= Ae^{-2t},$$

which—by equating the coefficients of $e^{-2t}$—yields that $A = 1$. Hence $Y(t) = te^{-2t}$ is a particular solution.

Now we must find the general solution $\phi$ to $\phi' + 2\phi = 0$. This is a simple, separable equation, and so by simple manipulation, we get $\phi'/\phi = -2$, i.e. $\ln|\phi| = -2t + c$ and finally $\phi(t) = Ce^{-2t}$ for some constant $C$. Thus our solution is

$$y(t) = \phi(t) + Y(t) = Ce^{-2t} + te^{-2t}.$$ 

The initial condition that $y(0) = 2$ requires that

$$2 = C \cdot 1 + 0 = C,$$

and so $y(t) = 2e^{-2t} + te^{-2t}$.

II.5.2. Example 11

$S(0) = -8000$, and $S' = .1S + k$ for some constant $k$:

8000 dollars is borrowed at an annual interest rate of 10%, compounded continuously with continuous payments at a rate of $k$.

1. Calculate $k$ if $S(3) = 0$.
2. Calculate the amount paid during the 3 year period.

Solution .:

The first step here is to solve the differential equation. We have two methods of doing this. For example, using an integrating factor:

$$S' - \frac{1}{10}S = k \iff e^{-t/10}S' - \frac{1}{10}e^{-t/10}S = ke^{-t/10},$$

so that $e^{-t/10}S = -10ke^{-t/10} + c$ for some $c$. Thus solving for $S$ yields

$$S(t) = -10k + ce^{t/10}.$$
Given the initial value of \( S(0) = -8000 \), we can say 
\[-8000 = S(0) = -10k + c \quad \implies c = 10k - 8000. \]

Hence \( S(t) \) is given by
\[ S(t) = -10k + (10k - 8000)e^{t/10}. \]

Solving the equation \( S(3) = 0 \) for \( k \) can be done as follows:
\[ 0 = -10k + 10ke^{3/10} - 8000e^{3/10} \quad \implies 8000e^{3/10} = k(10e^{3/10} - 10) \]
\[ \implies k = \frac{800e^{3/10}}{e^{3/10} - 1} \approx 3086.64. \]

If this is \( k \), then the student has been paying \( k \) for three years, meaning that
\[ \int_0^3 k \, dt = k \cdot 3 = \frac{2400e^{3/10}}{e^{3/10} - 1} \]
was paid in order to pay off the loan. How much of this was interest? Well this amount minus the 8000 of the original loan, i.e.
\[ 3k - 8000 = \frac{2400e^{3/10} - 8000e^{3/10} - 8000}{e^{3/10} - 1} = \frac{8000 - 5600e^{3/10}}{e^{3/10} - 1} \approx 1259.91. \]

### 11.5.3. Example 12

\( S(0) = -150,000 \), and \( S' = \frac{.06}{12} S + (800 + 10t) \):

150,000 dollars is borrowed at an annual interest rate of 6%, compounded continuously with continuous payments at a rate of 800 + 10t per month.

a. When is \( S(t) = 0 \)?

b. How large of a loan could be paid off in exactly 20 years given the payment of 800 + 10t per month?

**Solution:**

Again, the first step is to solve the differential equation. The setup here requires us to convert either the annual interest rate into monthly interest rate, or the monthly payments into yearly payments. Given that the annual interest rate is constant, we can just divide by 12. Note that \(.06/12 = 1/200\).

Here the differential equation is no longer separable, so we need to use an integrating factor, in this case \( e^{-0.6/12t} = e^{-t/200} \):

\[ (e^{-t/200} S)' = (800 + 10t)e^{-t/200}. \]

So now we integrate and get that
\[ e^{-t/200} S(t) = \int (800 + 10t)e^{-t/200} \, dt. \]

Calculating this requires integration by parts: the statement that \( \int f'g \, dt = fg - \int fg' \, dt \).

\[ f' = e^{-t/200} \quad f = -200e^{-t/200} \]
\[ g' = 10 \quad g = 800 + 10t \]

Thus integration by parts yields that
\[ \int (800 + 10t)e^{-t/200} \, dt = \int f'g \, dt = fg - \int fg' \, dt \]
\[ = -200(800 + 10t)e^{-t/200} - \int 10 \cdot (-200) - e^{-t/200} \, dt \]
\[ = -200(800 + 10t)e^{-t/200} + 560 \cdot 200 \cdot e^{-t/200} + c. \]
\[ = -2,000e^{-t/200} - 560,000e^{-t/200} + c \]
Solving the differential equation.

\[ S(t) = -2,000e^{-t/200} + c \]
\[ 0 = -2,000 \cdot 20 - 560,000 + (S_0 + 560,000)e^{240/200} \]
\[ \iff S_0 = (2,000 \cdot 20 + 560,000)e^{-240/200} - 560,000 \approx -246,758.02. \]
Hence the largest loan that could be paid off would be roughly 246,758.02 dollars.

A more complicated setup would be to convert the monthly to year. Given the earlier setup, this is like defining a function \( g(t) = y(12t) \). Using the chain rule, this means that \( g \) would need to satisfy the differential equation
\[ g'(t) = 12y'(12t) = .06g + 12(800 + 10 \cdot 12t), \]
which is slightly more complicated, but still gives you the same result—albeit in years instead of months.

\[ \text{II.5} \cdot 4. \quad \text{Example 13} \]

Suppose that \( u' = -\alpha u^4 \) with \( u(0) = 2000 \), and \( \alpha = 2 \cdot 10^{-12} \).
1. Solve the differential equation.
2. Find the time \( \tau \) where \( u(\tau) = 600 \).

**Solution:**

This is a separable differential equation that can be solved by dividing and integrating:

\[ \int \frac{u'}{u^4} \, dt = \int -\alpha \, dt \iff -\frac{1}{3u^3} = -\alpha t + c \iff u(t) = \frac{1}{\sqrt[3]{3\alpha t + C}}. \]

Given that \( u(0) = 2000 \), we can solve for \( C \):

\[ 2000 = \frac{1}{C^{1/3}} \iff C = \frac{1}{2000^3} = \frac{1}{8 \cdot 10^{-9}}. \]
\[ \implies u(t) = \frac{1}{\sqrt[3]{3\alpha t + 10^{-9}/8}}. \]

Using this, we can solve for \( \tau \) where \( u(\tau) = 600 \):

\[ 600 = \frac{1}{\sqrt[3]{3\alpha \tau + 10^{-9}/8}} \iff 3\alpha \tau + \frac{1}{8 \cdot 10^9} = \frac{1}{600^3} = \frac{1}{216 \cdot 10^6} \]
\[ \iff \tau = \frac{\frac{1}{216 \cdot 10^6} - \frac{1}{8 \cdot 10^9}}{3\alpha} = \frac{121625}{162} = 750.77, \]
where \( \tau \) is measured in seconds.
II.5 • 6. Example 15

Consider the differential equation $(t - 3) y' + y \ln t = 2t$ where $y(1) = 2$. Determine an interval where the solution this initial value problem is certain to exist.

Solution :

To put the equation in the proper form, we get

$$y' + \frac{\ln t}{t - 3} y = \frac{2t}{t - 3}.$$ 

Note that $p(t) = (\ln t)/(t - 3)$ is continuous whenever $t \neq 0$ and $t \neq 3$. Similarly, $2t/(t - 3)$ is continuous whenever $t \neq 3$. Given that $y$ is defined at 1, this means that $y$ is guaranteed to exist by the existence and uniqueness theorems whenever $t$ is (strictly) between 0 and 3.

II.5 • 7. Example 16

Consider the differential equation $y' = (t^2 + y^2)^{3/2}$.

State where in the $ty$-plane the hypotheses of The Existence and Uniqueness Theorem (II.1.E • 1) are satisfied.

Solution :

...
We’re given that \( f(t, y) = (t^2 + y^2)^{3/2} \). This is continuous everywhere. The partial derivative

\[
\frac{\partial}{\partial y} f(t, y) = f_y(t, y) = 3y(t^2 + y^2)^{1/2}
\]

is also continuous everywhere. Hence the hypotheses are satisfied everywhere.

---

**II.5 • 8. Example 17**

a. Verify that both \( y_1(t) = 1 - t \) and \( y_2(t) = -t^2/4 \) are solutions of the initial value problem

\[
y' = -t + \frac{\sqrt{t^2 + 4y}}{2}, \quad y(2) = -1.
\]

Where are these solutions valid?

b. Explain why the existence of two solutions of the given problem does not contradict the uniqueness part of The Existence and Uniqueness Theorem (II.1.E • 1).

c. Show that \( y = ct + c^2 \), where \( c \) is an arbitrary constant, satisfies the differential equation in part a for \( t \geq -2c \). If \( c = -1 \), the initial condition is also satisfied, and the solution \( y = y_1(t) \) is obtained. Show that there is no choice of \( c \) that gives the second solution \( y_2(t) \).

**Solution .:.**

a. You can just plug these in requiring that \( t^2 + 4y \geq 0 \).

b. Note that the function

\[
f(t, y) = -t + \frac{\sqrt{t^2 + 4y}}{2}, \quad f_y(t, y) = \frac{2}{\sqrt{t^2 + 4y}}
\]

has \( f \) as continuous for \( y \geq -t^2/4 \), which is true for \( y = -1 \) and \( t = 2 \). But \( f_y(t, y) \) is not continuous there, since \( \sqrt{t^2 + 4y} = 0 \) in that case. Hence the hypotheses of the theorem aren’t satisfied, and the consequences aren’t guaranteed.

c. You can just plug this in, and solve for \( t^2 + 4y(t) \geq 0 \). No value of \( c \) gives \( y_2 \), since there’s no \( t^2 \) term.

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**II.5 • 9. Example 18**

Suppose \( y' = -\beta y \), and \( x' = -\alpha xy \). \( x \) is the proportion of susceptibles, \( y \) is the proportion of carriers.

a. Solve for \( y(t) \) given that \( y(0) = y_0 \).

b. Solve for \( x(t) \) given that \( x(0) = x_0 \) using \( y(t) \) from part a.

c. Find the proportion of the population that escapes the epidemic by finding the limiting value of \( x \) as \( t \to \infty \).

**Solution .:.**

a. This is a simple exercise: \( y(t) = y_0 e^{-\beta t} \).

b. Using this, \( x' / x = -\alpha y_0 e^{-\beta t} \), so by integration, \( \ln |x| = (\alpha y_0 / \beta) e^{-\beta t} + c \), i.e.

\[
x(t) = Ce^{\alpha y_0 e^{-\beta t}/\beta}.
\]

Hence if \( x(0) = x_0 \), then \( x_0 = Ce^{\alpha y_0 / \beta} \), i.e.

\[
x(t) = x_0 e^{\alpha y_0 / \beta} (e^{-\beta t} - 1).
\]

c. The limit of \( x \) as \( t \to \infty \) is then \( x_0 e^{-\alpha y_0 / \beta} \).

---

**II.5 • 10. Example 19**

Consider \( y' = e^{2x} + y - 1 \). Find an integrating factor and solve the given equation.

**Solution .:.**
Write this in the usual form $My' + N = 0,$
\[ y' - e^{2x} - y + 1 = 0, \]
so that $M(x, y) = 1,$ and $N(x, y) = -e^{2x} - y + 1.$ Note that
\[ N_y = -1, \quad M_x = 0, \]
so that $\frac{N_y - N_x}{M} = -1$ is a function of just $t$ (and really just a constant). Hence solving $\mu' = -\mu$ gives an integrating factor. This of course yields $\mu = Ce^{-x},$ so that $\mu = e^{-x}$ works, ending up with
\[ e^{-x}y' - e^{-x}y - e^x + e^{-x} = 0. \] (II.1)
Note that this equation is exact, since for $W(x, y) = e^{-x},$ and $V(x, y) = -e^{-x}y - e^x + e^{-x},$
\[ W_x(x, y) = -e^{-x} = V_y(x, y). \]
Hence we can find a suitable $F$ where
\[ F'(y, t) = F_y' + F_x = e^{-x}y' - e^{-x}y - e^x - e^{-x}. \]
In particular, integrating with respect to $y$ of $W(x, y) = e^{-x}$ yields
\[ \int W(x, y) \, dy = e^{-x}y + c(x). \]
Integrating with respect to $x$ of $V(x, y) = -e^{-x}y - e^x + e^{-x}$ yields
\[ c(y) = \int V(x, y) \, dx - \int W(x, y) \, dy = \left[ e^{-x}y - e^x - e^{-x} \right] - e^{-x}y = -e^x - e^{-x}. \]
Hence $F(x, y) = e^{-x}y - e^x - e^{-x}$ works. This means that we simply must solve the equation
\[ e^{-x}y - e^x - e^{-x} = c \]
for $c$ a constant. This can be done so that $y = ce^x + e^{2x} + 1.$

---

**II.5 • 11. Example 20**

**Solution:**

Solve the equation
\[ \frac{dy}{dt} + (1 + 2t)y = 3t, \quad y(0) = 1. \]

We want an integrating factor $\mu$ satisfying
\[ \mu' = (1 + 2t)\mu, \]
meaning $\mu(t) = Ce^{t^2 + t}$ for some constant $C.$ Any $C \neq 0$ will work, so the simple choice of $C = 1$ works. So let $\mu(t) = e^{t^2 + t}.$ By our choice of $\mu,$
\[ (\mu y)' = \mu y' + (1 + 2t)\mu y = 3t\mu. \]
So integrating and dividing by $\mu$ yields that
\[ y(t) = \frac{1}{\mu(t)} \int 3t\mu(t) \, dt = e^{-t^2 - t} \int 3te^{t^2 + t} \, dt. \]
This integral can be simplified slightly, but we cannot remove the integral. So let $F(t) = \int_0^t 3se^{s^2 + s} \, ds.$ Thus for some $C,$
\[ y(t) = e^{-t^2 - t}(F(t) + C). \]
The requirement that $y(0) = 1$ says that $1 = F(0) + C,$ i.e. $C = 1 - F(0).$ Hence we can write
\[ y(t) = e^{-t^2 - t}(F(t) + 1 - F(0)). \]
II.5 • 12. Example 21

(a) Find the critical points of the equation
\[ \frac{dy}{dt} = y \cdot (2y - 1) \cdot (y - 2). \]

(b) Find the type of each critical point (stable, unstable, semi-stable).

(c) If \( y(0) = 1 \) what is \( \lim_{t \to \infty} y(t) \)?
If \( y(0) = 1/4 \) what is \( \lim_{t \to \infty} y(t) \)?
If \( y(0) = 3/2 \) what is \( \lim_{t \to \infty} y(t) \)?

**Solution:**

(a) The critical points of an autonomous differential equation \( y' = f(y) \) are just the values \( a \) where \( f(a) = 0 \). These give constant solutions \( y = a \) to the differential equation. In this case, \( f(y) = y(2y - 1)(y - 2) \), which is 0 iff a factor is zero, i.e. \( y = 0, y = 1/2, \) or \( y = 2 \).

(b) For \( y < 0, f(y) < 0 \) (the product of three negative numbers).
   For \( 0 < y < 1/2, f(y) > 0 \) (the product of a positive, and two negative numbers).
   For \( 1/2 < y < 2, f(y) < 0 \) (the product of two positives, and a negative number).
   For \( 2 < y, f(y) > 0 \) (the product of three positive numbers).
   Hence when near 0, solutions move away from 0, meaning \( y = 0 \) is unstable.
   When near 1/2, solutions move towards 1/2, meaning \( y = 1/2 \) is stable.
   When near 2, solutions move away from 2, meaning \( y = 2 \) is stable.

(c) The initial value \( y(0) = 1 \) is between 1/2 and 2, and so \( y(t) \) tends to 1/2.
   The initial value of \( y(0) = 1/4 \) is between 0 and 1/2, and so \( y(t) \) tends to 1/2.
   The initial value of \( y(0) = 3/2 \) is between 1/2 and 2, and so \( y(t) \) tends to 1/2.

II.5 • 13. Example 22

1. Find the critical points of the equation \( y' = y(1 + y^2 - 5y) \).
2. Find the type of each critical point.
3. If \( y(0) = 1 \), what is \( \lim_{t \to \infty} y(t) \)?

**Solution:**

1. The critical points are when \( y(1 + y^2 - 5y) = 0 \), which happens whenever \( y = 0 \), or \( 1 + y^2 - 5y = 0 \). The latter happens iff \( y = (5 \pm \sqrt{21})/2 \). So the critical points are \( 0, (5 - \sqrt{21})/2 \), and \( (5 + \sqrt{21})/2 \). Call these latter two \( \lambda_1 \) and \( \lambda_2 \) so that our points are \( 0 < \lambda_1 < \lambda_2 \).

2. Note that we can factor
\[ y' = y \cdot (y - \lambda_1)(y - \lambda_2). \]
   Note also that \( \sqrt{21} < 5 \), since \( \sqrt{21}^2 = 21 < 25 = 5^2 \). As a result, \( (5 - \sqrt{21})/2 > 0 \) so that \( \lambda_1 \) and \( \lambda_2 \) are both positive. Hence if \( y < 0 \), then \( y' < 0 \). If \( 0 < y < \lambda_1 \), then \( y' > 0 \). This means that 0 is unstable.
   If \( \lambda_1 < y < \lambda_2 \), then \( y' < 0 \). Since we know \( 0 < y < \lambda_1 \) implies \( y' > 0 \), \( \lambda_1 \) is stable.
   If \( \lambda_2 < y \), then \( y' > 0 \). Since we know \( \lambda_1 < y < \lambda_2 \) implies \( y' < 0 \), \( \lambda_2 \) is unstable.

3. We need to see where 1 is between 0, \( \lambda_1 \), \( \lambda_2 \), and \( \infty \). Note that \( \lambda_1 < 1 \). To see this, note that \( \sqrt{21} > 3 \). Hence \( 5 - \sqrt{21} < 5 - 3 = 2 \) so that \( (5 - \sqrt{21})/2 < 1 \). On the other hand, it’s clear that \( \lambda_2 > 1 \), since \( 1 < 5/2 < \lambda_2 \). Therefore \( \lambda_1 < y(0) < \lambda_2 \) and by the stability of \( \lambda_1 \) above, it follows that \( \lim_{t \to \infty} y(t) = \lambda_1 \).
Solve the equation
\[ y' = \left(1 + \frac{1}{t}\right)y + \begin{cases} 1 & \text{if } 1 \leq t \leq 2 \\ 2 & \text{if } 2 < t, \end{cases} \]
with initial condition \( y(1) = 1 \).

**Solution:**

Proceed by integrating factors: we want a \( \mu \) that satisfies
\[ \mu' = -\left(1 + \frac{1}{t}\right)\mu. \]

By dividing by \( \mu \) and integrating, this yields that \( \mu \) will be of the form, noting that \(|t| = t\) since \( t > 0 \),
\[ \ln |\mu| = -t - \ln |t| + c \leftrightarrow \mu = Ce^{-t} \frac{1}{|t|} = \frac{C}{te^t}. \]
So we can choose \( \mu = 1/(te^t) \) since any choice of \( C \) will work. Therefore
\[ \mu y' + \mu \left(1 + \frac{1}{t}\right)y = \begin{cases} 1/(te^t) & \text{if } 1 \leq t \leq 2 \\ 2/(te^t) & \text{if } 2 < t, \end{cases} \]
\[ \mu y' + \mu y = \mu = Ce^{-t}. \]

To integrate the right hand side, we need to consider two cases, in essence separating the integral over two intervals. Let
\[ g(t) := \begin{cases} 1/(te^t) & \text{if } 1 \leq t \leq 2 \\ 2/(te^t) & \text{if } 2 < t. \end{cases} \]
Suppose \( 1 \leq t \leq 2 \). In this case, \( g(t) \) is just \( 1/(te^t) \), whose integral can’t be done by hand. So let \( E(t) = \int_1^t 1/(xe^{x^2}) \, dx \). In the case that \( t > 2 \), then we need to split the integral:
\[ \int_1^t g(x) \, dx = \int_1^2 g(x) \, dx + \int_2^t g(x) \, dx = E(2) + \int_2^t \frac{2}{xe^{x^2}} \, dx = E(2) - 2E(t) \]
Thus we can write
\[ \int_1^t g(x) \, dx = \begin{cases} E(t) & \text{if } 1 \leq t \leq 2 \\ 2E(t) - E(2) & \text{if } 2 < t. \end{cases} \]

Returning to the point of this integral, we get
\[ \int_1^t (\mu(x)y(x))' \, dx = \mu(t)y(t) - \mu(1)y(1) = \int_1^t g(x) \, dx, \]
and thus we can solve for \( y(t) \): note that \( \mu(1) = 1/e \), and \( y(1) = 1 \) so that
\[ y(t) = \frac{\mu(1)y(1)}{\mu(t)} + \frac{1}{\mu(t)} \int_1^t g(x) \, dx = te^{-t} + \begin{cases} te^tE(t) & \text{if } 1 \leq t \leq 2 \\ 2te^tE(t) - E(2)te^t & \text{if } 2 < t. \end{cases} \]
Chapter III. Discrete Problems

Section III.1. First-order difference equations

Let’s take a break from looking at continuous functions to looking at discrete problems. In particular, sequences \( \langle y_0, y_1, y_2, \cdots \rangle \) defined by recursion, meaning that we get the next value from the previous ones:

\[
y_{n+1} = f(n, y_n),
\]

for some function \( f \). This is supposed to be analogous to the differential equation \( y' = f(t, y) \). Now clearly given an initial value \( y_0 \), the above equation uniquely defines a function: we just compute \( f(0, y_0) \) to get \( y_1 \), and then we compute \( f(1, y_1) \) to get \( y_2 \), \( f(2, y_2) \) for \( y_3 \), and so on. But this process is both lengthy, and uninformative about the general trends of solutions.

The interest with these equations \( y_{n+1} = f(n, y_n) \) is giving a closed form solution, meaning an expression that calculates \( y_n \) at any \( n \) (given the initial value \( y_0 \)). This would allow us to see what happens to \( y_n \) as \( n \to \infty \), for instance. If we can still find a closed form solution for arbitrary \( y_0 \), we can also see what happens if we vary the initial value \( y_0 \). In general, however, finding such closed form solutions is either impossible, or difficult at the very least. Although we can calculate any given value just by calculating \( f \) a bunch of times, it’s hard to figure out how to calculate just based on \( n \) without knowing previous values.

There is no sure-fire way to find a closed form solution. Generally, the best method is to calculate the first several values in terms of \( y_0 \), and then try to generalize. Of course, you then want to confirm that your solution \( g(n) \) is actually correct by plugging it into \( f \), and making sure \( f(n, g(n)) \) is just \( g(n + 1) \) for all \( n \).

Now there are more analogous properties of difference equations and differential equations\(^\ast\). For example, like with autonomous differential equations, we have a notion of stability of constant solutions (solutions where \( y_n = c \) for all \( n \), i.e. the sequence \( \langle y_0, y_1, y_2, \cdots \rangle = \langle c, c, c, \cdots \rangle \)).

### III.1.1. Definition 8

Consider the first-order difference equation \( y_{n+1} = f(y_n) \) for some \( f \). The constant solution or equilibrium \( c \) is

1. **stable** iff solutions with \( y_0 \) “near” \( c \) approach \( c \) as \( t \to \infty \).
2. **unstable** iff solutions with \( y_0 \) “near” \( c \) diverge from \( c \) as \( t \to \infty \).

However, being “near” \( c \) is vague, even more-so than in autonomous differential equations (equations of the form \( y' = f(y) \) for some \( f \)). In many cases it may be unclear where a solution is stable.

Section III.2. Euler’s Method

### § III.2.A. The method

\(^\ast\)Well, really it’s the definitions that are analogous.
Euler’s method is a pretty intuitive way of approximating solutions. The idea is a step-by-step process where you look at what direction you’re supposed to go, and then go a few steps in that direction, and then re-evaluate what direction you’re supposed to go, and so on. The direction is determined by \( y' \), and the steps you take are arbitrary. But intuitively, if you keep correcting your direction really frequently—taking only small steps—you’ll be more accurate in the end. This unfortunately means that if you want an accurate approximation, you’ll need to take a lot of time to do the process over and over many times.

It’s usually the case that the step size is the same every time: some value \( h \). If you step by \( h \) every time, then you’re approximating the value of the function at \( y_0 = y(t_0), y_1 = y(t_0 + h), y_2 = y(t_0 + 2h), \) and so on: \( y_n = y(t_0 + nh) \). This creates a first-order difference equation.

### III.2.A • 1. Result 15

Suppose \( y' = f(t, y) \) and \( y(t_0) = y_0 \). Thus Euler’s method gives the first-order difference equation

\[
y_{n+1} = y_0 + h \cdot f(t_0 + nh, y_n),
\]

which is supposed to approximate the solution to the differential equation at \( t_0 + nh \):

\[
y(t_0 + nh) \approx y_n.
\]

The resulting table of values of \((t_1, y_1), (t_2, y_2), \ldots\) yields an approximation of the solution to \( y' = f(y, t) \) with initial value \( y(t_0) = y_0 \). Again, this process is long if done by hand, so such problems almost need a calculator if you don’t want to waste your time.

### § III.2.B. Errors in Euler’s method

Recall that Euler’s method is a step-by-step process where you look at where you’re supposed to go, then go a few steps in that direction, and then re-evaluate what direction you’re supposed to go, over and over. Now if we can actually solve the initial value problem \( y' = f(t, y) \) with \( y(t_0) = y_0 \) for \( y \), then we can look at the difference between the approximated values, and the actual values: the error in our approximations. Even better, if we can get a closed form solution to the difference equation in Result 15 (III.2.A • 1), then we can look at how our error develops over time, and how good our approximated solution becomes.

### III.2.B • 1. Definition 9

Consider the differential equation \( y' = f(t, y) \) with \( y(t_0) = y_0 \).

Let \( \phi(t) \) be a solution this.

Let \( \langle y_0, y_1, y_2, \cdots \rangle \) be a sequence approximating \( \phi \) by Euler’s method with step size \( h \).

The error or global truncation error of this approximation is the sequence \( \langle E_0, E_1, E_2, \cdots \rangle \) where

\[
E_n := \phi(t_n) - y_n.
\]

The local truncation error of the method is the sequence \( \langle e_0, e_1, e_2, \cdots \rangle \) would be the errors of the next step in Euler’s method if we correctly started with \( \phi(t_{n-1}) \):

\[
e_n := \phi(t_n) - \left[ \phi(t_{n-1}) + h \cdot f(t_n, \phi(t_{n-1})) \right],
\]

\[
E_n := \phi(t_n) - \left[ y_{n-1} + h \cdot f(t_n, y_{n-1}) \right].
\]

Errors in \( E_n \) have accumulated, but the errors on \( e_n \) have not. Given that \( y_n \approx \phi(t_n) \) for all \( n \), we should have that \( e_n \approx E_n \) (as the above would suggest), but this is often not the case, especially for large \( n \). The book[1] has its own favorite way of calculating the local truncation error with the following.
### III.2.B. Definition 10

Consider the differential equation \( y' = f(t, y) \) with \( y(t_0) = y_0 \).
Let \( \phi(t) \) be a solution to this with continuous second derivative \( \phi'' \).
Consider Euler’s method with step size \( h \): \( t_n = t + nh \).
The local truncation error of this method is the sequence \( \langle e_0, e_1, e_2, \cdots \rangle \) where
\[
e_n := \frac{1}{2} \phi''(\bar{t})h^2,
\]
where \( \bar{t} \) is a fixed, unknown (to us) value between \( t_n \) and \( t_{n+1} \).

The value of \( \bar{t} \) is one is given by a cut-off Taylor polynomial. In particular, we approximate \( \phi \) with its Taylor series, stopping at \( \phi'' \). This isn’t exactly \( \phi \), however, so we allow the \( \phi'' \) term to be evaluated somewhere between \( t_n \) and \( t_{n+1} \) to ensure the equality actually holds:
\[
\phi(t_{n+1}) = \phi(t_n) + \phi'(t_n)h + \frac{1}{2} \phi''(\bar{t})h^2.
\]
This can be done so long as \( \phi'' \) is continuous between \( t_n \) and \( t_{n+1} \).

### §III.2.C. Improvements in Euler’s Method

The improved Euler method is to approximate based on two things: where you are, and where you would end up. This requires solving an equation to get your result, but the approximation will be better. In particular,

### III.2.C. Definition 11

Consider the differential equation \( y' = f(t, y) \) for some function \( f \) and initial value \( y(t_0) = y_0 \).
Let \( t_n = t_0 + nh \) for some fixed \( h \).
The improved Euler method with step-size \( h \) is a sequence \( \langle y_0, y_1, \cdots \rangle \) satisfying
\[
y_{n+1} = y_n + h \frac{f_n + f(t_{n+1}, y_n + hf_n)}{2}, \quad \text{where} \ f_n := f(t_n, y_n).
\]
The sequence is supposed to approximate the actual solution \( \phi \) at the points \( t_0, t_1, \) and so on:
\[
y_0 = \phi(t_0)
y_1 \approx \phi(t_1)
y_2 \approx \phi(t_2)
y_3 \approx \phi(t_3)
\]
\[\vdots\]
Note that \( h \) and \( t_0 \) determine \( t_n \) for every \( n \), so we can’t skip values in our calculations. If \( t_0 = 0 \) and \( h = .025 \), to approximate \( \phi(25) \), we would have to calculate \( t_1, t_2, \cdots, \) up to \( t_{1000} \).

In essence, the improved Euler method is similar to using trapezoids in Riemann integration as opposed to rectangles.
Section III.3. Worked-Out Examples

III 3.1. Example 24

Consider the differential equation with initial condition $y(0) = 1$

$$y' = 2y - 1.$$  

a. Approximate $y$ at $t = .1, .2, .3,$ and $.4$ using Euler’s method with $h = .1$.
b. Approximate $y$ at those same values with $h = .05$.
c. Approximate $y$ at those same values with $h = .025$.
d. Find the solution to the differential equation, and compute its values at $.1, .2, .3,$ and $.4$, comparing with the approximations.

Solution:

a. We get the following table of values using Euler’s method with $h = .1$, and $f(t, n) = 2y - 1$, stopping once we’ve gone over $t = .4$:

<table>
<thead>
<tr>
<th>$t_n$</th>
<th>$f(t_n, y_n)$</th>
<th>$y_n$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>.1</td>
<td>1.2</td>
<td>1.1</td>
</tr>
<tr>
<td>.2</td>
<td>1.44</td>
<td>1.22</td>
</tr>
<tr>
<td>.3</td>
<td>1.728</td>
<td>1.364</td>
</tr>
<tr>
<td>.4</td>
<td>2.0736</td>
<td>1.5368</td>
</tr>
</tbody>
</table>

b. We get the following table of values using Euler’s method with $h = .05$, stopping once we’ve gone over $t = .4$.

<table>
<thead>
<tr>
<th>$t_n$</th>
<th>$f(t_n, y_n)$</th>
<th>$y_n$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>.05</td>
<td>1.1</td>
<td>1.05</td>
</tr>
<tr>
<td>.1</td>
<td>1.21</td>
<td>1.105</td>
</tr>
<tr>
<td>.15</td>
<td>1.331</td>
<td>1.1655</td>
</tr>
<tr>
<td>.2</td>
<td>1.4641</td>
<td>1.2321</td>
</tr>
<tr>
<td>.25</td>
<td>1.6105</td>
<td>1.3053</td>
</tr>
<tr>
<td>.3</td>
<td>1.7716</td>
<td>1.3858</td>
</tr>
<tr>
<td>.35</td>
<td>1.9487</td>
<td>1.4743</td>
</tr>
<tr>
<td>.4</td>
<td>2.1436</td>
<td>1.5718</td>
</tr>
</tbody>
</table>

c. We get the following table of values using Euler’s method with $h = .025$, stopping once we’ve gone over $t = .4$. These values are cut-off and rounded for readability.

<table>
<thead>
<tr>
<th>$t_n$</th>
<th>$f(t_n, y_n)$</th>
<th>$y_n$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>.025</td>
<td>1.05</td>
<td>1.025</td>
</tr>
<tr>
<td>.05</td>
<td>1.1025</td>
<td>1.0513</td>
</tr>
<tr>
<td>.075</td>
<td>1.1576</td>
<td>1.0788</td>
</tr>
<tr>
<td>.1</td>
<td>1.2155</td>
<td>1.1078</td>
</tr>
<tr>
<td>.125</td>
<td>1.2763</td>
<td>1.1381</td>
</tr>
<tr>
<td>.15</td>
<td>1.3401</td>
<td>1.1700</td>
</tr>
<tr>
<td>.175</td>
<td>1.4071</td>
<td>1.2036</td>
</tr>
<tr>
<td>.2</td>
<td>1.4775</td>
<td>1.2387</td>
</tr>
</tbody>
</table>
It

<table>
<thead>
<tr>
<th>$t_n$</th>
<th>$f(t_n, y_n)$</th>
<th>$y_n$</th>
</tr>
</thead>
<tbody>
<tr>
<td>.225</td>
<td>1.5513</td>
<td>1.2757</td>
</tr>
<tr>
<td>.25</td>
<td>1.6289</td>
<td>1.3144</td>
</tr>
<tr>
<td>.275</td>
<td>1.7103</td>
<td>1.3552</td>
</tr>
<tr>
<td>.3</td>
<td>1.7959</td>
<td>1.3979</td>
</tr>
<tr>
<td>.325</td>
<td>1.8856</td>
<td>1.4428</td>
</tr>
<tr>
<td>.35</td>
<td>1.9799</td>
<td>1.4900</td>
</tr>
<tr>
<td>.375</td>
<td>2.0789</td>
<td>1.5395</td>
</tr>
<tr>
<td>.4</td>
<td>2.1829</td>
<td>1.5914</td>
</tr>
</tbody>
</table>

\[ \phi(t) = \frac{1}{2} + \frac{1}{2}e^{2t} \]

We then have the following table of values, comparing the different approximations to the actual solution at $t = .1, .2, .3, \text{ and } .4$.

<table>
<thead>
<tr>
<th>$t_n$</th>
<th>$\phi$</th>
<th>$h = .025$</th>
<th>$h = .05$</th>
<th>$h = .1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>.1</td>
<td>1.1107</td>
<td>1.1078</td>
<td>1.105</td>
<td>1.1</td>
</tr>
<tr>
<td>.2</td>
<td>1.2459</td>
<td>1.2387</td>
<td>1.2321</td>
<td>1.22</td>
</tr>
<tr>
<td>.3</td>
<td>1.4111</td>
<td>1.3979</td>
<td>1.3858</td>
<td>1.364</td>
</tr>
<tr>
<td>.4</td>
<td>1.6128</td>
<td>1.5914</td>
<td>1.5718</td>
<td>1.5368</td>
</tr>
</tbody>
</table>

So we are underestimating $\phi$, and with larger $h$, the difference becomes worse and worse.

### III.3 • 2. Example 25

Solve the difference equation \[ y_{n+1} = \frac{9}{10} y_n \]
in terms of $y_0$, and describe the behavior of the solution as $n \to \infty$.

**Solution**: The solution is that $y_n = (-9/10)^n y_0$. This can be proven by induction: clearly for $y_0$ this holds. If it holds for $n$, then $y_{n+1} = (-9/10)y_n = (-9/10)^{n+1}y_0$, exactly as predicted.

Note that since $|-9/10| = 9/10 < 1$, the limit $\lim_{n \to \infty} (9/10)^n = 0$, meaning $\lim_{n \to \infty} y_n = 0$.

### III.3 • 3. Example 26

Solve the difference equation \[ y_{n+1} = (-1)^{n+1} y_n \]
in terms of $y_0$, and describe the behavior of the solution as $n \to \infty$.

**Solution**: The solution is that $y_n = (-1)^{(n^2-n)/2} y_0$. This can be proven by induction: clearly for $y_0$ this holds. Suppose it holds for $n$. We know that

\[
\frac{m^2 - m}{2} + m + 1 = \frac{(m + 1)^2 - (m + 1)}{2}
\]

for all $m$. So applying this to $m = n$ and using the inductive hypothesis yields

\[ y_{n+1} = (-1)^{n+1}y_n = (-1)^{n+1+\frac{2-n^2}{2}}y_0 = (-1)\frac{(n+1)^2-(m+1)}{2}y_0, \]

yielding the equality. This solution has no limit as $n \to \infty$ as it will go back and forth between $y_0$ and $-y_0$. In particular, it will start with $y_0$, and then go $-y_0, y_0, y_0, y_0 \text{ over and over}$.
Example 27

Solve the difference equation

\[ y_{n+1} = \frac{1}{2} y_n + 6 \]

in terms of \( y_0 \), and describe the behavior of the solution as \( n \to \infty \).

Solution :
Calculating the first few terms in terms of \( y_0 \) yields

\[
\begin{align*}
y_1 &= \frac{1}{2} y_0 + 6 \\
y_2 &= \frac{1}{4} y_0 + \frac{1}{2} 6 + 6 \\
y_3 &= \frac{1}{8} y_0 + \frac{1}{4} 6 + \frac{1}{2} 6 + 6 \\
y_4 &= \frac{1}{16} y_0 + \frac{1}{8} 6 + \frac{1}{4} 6 + \frac{1}{2} 6 + 6.
\end{align*}
\]

This suggests, but does not demonstrate, that for \( n \geq 1 \),

\[
y_n = \frac{y_0}{2^n} + \sum_{k=0}^{n-1} \frac{6}{2^k} = \frac{y_0}{2^n} + 6 \cdot \frac{(1/2)^{n-1+1} - 1}{(1/2) - 1} = \frac{y_0}{2^n} + 6(2 - 2^{1-n}).
\]

Proving this is a matter of induction, and is not very difficult. Clearly as \( n \to \infty \), \( y_0/2^n \) goes to 0. The term on the right then goes to 6 times 2. Hence \( \lim_{n \to \infty} y_n = 12 \).

Example 28

Suppose \( y_0 = -100,000 \) with \( y_{n+1} \) for \( n \geq 0 \) calculated by

\[
y_{n+1} = \left( 1 + \frac{9}{100} \right) \left( 1 - \frac{1}{12} \right) y_n + p
\]

for some constant \( p \) (here \( n \) in months).

1. What \( p \) (monthly payment) has \( y_{30-12} = 0 \)?
2. What \( p \) has \( y_{20-12} = 0 \)?
3. What is the total amount paid in each case?

Solution :
We have to work in months, since the payments are paid over months rather than in one lump sum at the end of the year. And these differences change the calculations over the rest of the months. So there’s no nice way to change from months to years.

The first step here is to get a closed form solution for \( y_n \) in terms of \( p \). Write \( A \) for 1209/1200 for the sake of space. Calculating the first few terms in terms of \( y_0 \) yields

\[
\begin{align*}
y_1 &= Ay_0 + p \\
y_2 &= A^2 y_0 + Ap + p \\
y_3 &= A^3 y_0 + A^3 p + Ap + p \\
\end{align*}
\]
This suggests, but does not demonstrate, that for \( n \geq 1 \),
\[
y_n = A^n y_0 + \sum_{k=0}^{n-1} A^k p
\]
\[
= A^n y_0 + p \cdot \frac{A^n - 1}{A - 1}.
\]
Proving this is a simple proof by induction.

1. This assumption says that
\[
0 = A^{30-12} y_0 + p \cdot \frac{A^{30-12} - 1}{A - 1}
\]
solving for \( p \) is easy, as this is just a messy equation of the form \( c + pd = 0 \) for some constants \( c, d \). The result is that
\[
p = -\frac{y_0 A^{30-12} (A - 1)}{A^{30-12} - 1} \approx 804
\]

2. In a similar way as above,
\[
p = -\frac{y_0 A^{20-12} (A - 1)}{A^{20-12} - 1} \approx 900
\]

3. We can calculate the amount paid easily: if it took \( n \) steps to get to 0, we’ve paid \( n \cdot p \).

So if it took 30 years, or 30 \cdot 12 \) months, then we’ll pay the amount in 1. times 360, which is roughly 290,000.

If it took 20 years, or 20 \cdot 12 \) months, then we pay the amount in 2. times 240, which is roughly 215,000

---

### III.3.6. Example 29

Consider the differential equation
\[
y' = \frac{1}{2} - t + 2y
\]
with \( y(0) = 1 \).

Find approximate values of the solution at \( t = .5, 1, 1.5, \) and 2 using Euler’s method with \( h = .025 \).

**Solution**: Using a computer for this is almost mandatory, since one must calculate the value of \( y_n \) for \( 2/h = 80 \) different \( n \) before finding all four values. Skipping this, we get that the actual solution \( \phi \) is approximated as follows:

1. \( \phi(.5) \) is approximately \( y_{20} \approx 2.9033 \) (note that this is an approximation to the approximation, since the value of the approximation \( y_{20} \) has been cut-off and rounded).
2. \( \phi(1) \) is approximately \( y_{40} \approx 7.5400 \) (again, approximately the approximation).
3. \( \phi(1.5) \) is approximately \( y_{60} \approx 19.4292 \).
4. \( \phi(2) \) is approximately \( y_{80} \approx 50.5614 \).
Obtain a formula for the local truncation error for Euler’s method in terms of $t$ and the exact solution $y = \phi(t)$ for the differential equation
\[ y' = 5t - 3\sqrt{y}, \quad y(0) = 2. \]

**Solution:**

Note that you’re not supposed to actually find the exact solution $\phi(t)$. Instead, we merely need to calculate $\phi''$ in order to use the book’s definition[1] *Definition 10 (III.2.B • 2)*:

\[
\phi''(t) = \frac{d}{dt}(5t - 3\sqrt{\phi(t)}) = 5 - \frac{3}{2} \frac{1}{\sqrt{\phi(t)}} \phi'(t)
\]
\[
= 5 - \frac{3}{2} \frac{1}{\phi(t)} (5t - 3\sqrt{\phi(t)})
\]
\[
= 5 - \frac{15}{2} \frac{t}{2\sqrt{\phi(t)}} + \frac{9}{2}
\]
\[
= \frac{19}{2} - \frac{15}{2} \frac{t}{\sqrt{\phi(t)}}.
\]

So using *Definition 10 (III.2.B • 2)*, for some particular $\tilde{t}$ between $t + nh$ and $t + (n + 1)h$,

\[
e_n = \frac{1}{2} \phi''(\tilde{t}) h^2
\]
\[
= \frac{1}{2} \left( \frac{19}{2} - \frac{15}{2} \frac{\tilde{t}}{\sqrt{\phi(\tilde{t})}} \right) h^2 = \frac{1}{4} \left( 19 - 15\frac{\tilde{t}}{\phi(\tilde{t})} \right) h^2.
\]
Chapter IV. Second-Order, Linear, Differential Equations

First order equations had many different types that we studied, but here we will be focused on two types: homogeneous, and non-homogeneous equations, both of which are kinds of linear equations. The primary kinds of equations we will be working with will be of the form

\[ y'' + p(t)y' + q(t)y = g(t) \]

for functions \( p, q, \) and \( g \). First, we need to consider what happens when \( g \) is 0.

**Section IV.1. Homogeneous Equations**

There is little connection with “homogenous” used in differential equations, and the term used in the rest of mathematics. For our purposes, we have the following definition.

**IV.1 • 1. Definition 12**

A second-order differential equation is **homogeneous** iff it is of the form

\[ P(t)y'' + Q(t)y' + R(t)y = 0, \]

for some functions \( P, Q, \) and \( R \).

In other words, it is homogeneous iff it is linear with no constant (with respect to \( y \)) term. The solutions of these equations will be of the form \( C\phi_1 + D\phi_2 \) for solutions \( \phi_1 \) and \( \phi_2 \) where \( C \) and \( D \) are determined using the initial values of any given initial value problem: \( y_0 \) and \( y'_0 \). The expression \( C\phi_1 + D\phi_2 \) is a **linear combination**, a phrase that may be thrown around a lot at this point. We have the following definition.

**IV.1 • 2. Definition 13**

Let \( \phi_1, \phi_2, \cdots, \phi_n \) be some things (functions, vectors, etc.). A linear combination of these is anything of the form

\[ c_1\phi_1 + c_2\phi_2 + \cdots + c_n\phi_n \]

for constants \( c_1, c_2, \cdots, \) and \( c_n \), the **coefficients**.

Finding solutions can be more difficult, but the idea is that we only need to find two solutions to get the general solution. The key thing, however, is that these solutions must be different enough. What characterizes this is having a non-zero **Wronskian**.

**IV.1 • 3. Definition 14**

Let \( f \) and \( g \) be two differentiable functions. The **Wronskian** is the function given by

\[ W[f, g] = \det \begin{bmatrix} f & g \\ f' & g' \end{bmatrix} = fg' - f'g. \]
Consider the differential equation
\[ y'' + p(t)y' + q(t)y = 0, \quad y(t_0) = y_0, \quad y'(t_0) = y'_0, \]
for some \( p \) and \( q \). Suppose \( \phi_1 \) and \( \phi_2 \) are distinct solutions to this.

So long as \( W[\phi_1, \phi_2](t_0) \neq 0 \), any solution \( y \) around \( t_0 \) can be written as a linear combination of \( \phi_1 \) and \( \phi_2 \):
\[ y(t) = C\phi_1(t) + D\phi_2(t), \]
for constants \( C \) and \( D \). \( \phi_1 \) together with \( \phi_2 \) make up a fundamental set of solutions.

In some sense, we’re making sure that \( f \) and \( g \) are different enough: that the vector \( \langle f, f' \rangle \) isn’t just a scaled version of \( \langle g, g' \rangle \), and similarly that \( \langle f, g \rangle \) isn’t just a scaled version of \( \langle f', g' \rangle \). In another sense, we’re making sure that \( 0 \) isn’t a linear combination of these (when the coefficients are non-zero). Note the following theorem that tells us what \( W[\phi_1, \phi_2] \) is without having to find \( \phi_1 \) and \( \phi_2 \):

Consider the differential equation
\[ y'' + p(t)y' + q(t)y = 0, \]
and suppose \( \phi_1 \) and \( \phi_2 \) are solutions to this. Therefore
\[ W[\phi_1, \phi_2](t) = Ce^{-\int p(t) \, dt} \]
for some constant \( C \) (depending on \( \phi_1 \) and \( \phi_2 \)).

In particular, this means that if the Wronskian of two solutions is \( 0 \) somewhere on an interval, since \( e^{(\text{whatever})} \) is never \( 0 \), \( C \) must be \( 0 \). This would imply that it is always \( 0 \) on that interval, so that the inverse holds too: if the Wronskian is not \( 0 \) somewhere on an interval, then it is never \( 0 \) on that interval.

In the case that \( p \) and \( q \) are continuous everywhere, this means the Wronskian of two solutions is either always \( 0 \) or never \( 0 \). However, if \( p \) and \( q \) have discontinuities, the Wronskian may be non-0 just on a small interval, and \( 0 \) everywhere else. Regardless, we still get our desired equalities from previous theorems: on such an interval, any solution can be written as a linear combination of those two solutions.

Note also that Abel’s Theorem (IV.1 • 5) references the Wronskian of solutions to second-order, homogeneous, differential equations, not functions in general. For example, the Wronskian of \( e^{t} \) and \( \cos(t) \) is
\[ W[e^t, \sin t] = \det \begin{bmatrix} e^t & \sin t \\ e^t & \cos t \end{bmatrix} = e^t \cos t - e^t \sin t, \]
which is never constantly \( 0 \) on an interval\(^1\), but it is occasionally \( 0 \), e.g. at \( t = \pi/4 \).

Now returning back to equations, in the case that \( P, Q, \) and \( R \) are constants in the equation \( P(t)y'' + Q(t)y' + R(t)y = 0 \), we have the following result that allows us to characterize the solutions.

Consider the differential equation
\[ ay'' + by' + cy = 0. \]
If there are two solutions \( r_1 \neq r_2 \) (real or complex) to
\[ ar^2 + br + c = 0, \]
then all solutions of the differential equation are of the form
\[ Ce^{r_1t} + De^{r_2t} \]
for some constants \( C \) and \( D \).

\(^1\)at least on an interval that isn’t a single point
A hiccup, however, is when there is only one solution to $ar^2 + br + c = 0$, i.e. when using the quadratic formula, the discriminant is 0.

Consider the differential equation $ay'' + by' + cy = 0$. If $r = r_1 = r_2$ is the only solution to $ar^2 + br + c = 0$, then all solutions of the differential equation are of the form $Ce^{rt} + De^{rt}$, for some constants $C$ and $D$.

Proof :.

**Claim 1**

$\phi_1(t) = te^{rt}$ is a solution to $ay'' + by' + cy = 0$.

Proof :.

We can calculate that

$$\phi_1'(t) = rte^{rt} + e^{rt}, \quad \phi_1''(t) = r^2te^{rt} + 2re^{rt}.$$  

Hence

$$a\phi_1'' + b\phi_1' + c\phi_1 = ar^2te^{rt} + 2are^{rt} + brte^{rt} + be^{rt} + cte^{rt}$$

$$= te^{rt}(ar^2 + br + c) + e^{rt}(2ar + b)$$

$$= c + e^{rt}(2ar + b).$$

Since $r$ is the only solution to $ax^2 + bx + c = 0$, we can write

$$ax^2 + bx + c = a(x - r)^2 = ax^2 - 2arx + ar^2,$$

which then requires $b$ to be $-2ar$. Hence $2ar + b = 0$ so that above,

$$a\phi_1'' + b\phi_1' + c\phi_1 = 0 \dashrightarrow$$

It’s clear that $\phi_2(t) = e^{rt}$ is a solution to $ay'' + by' + cy = 0$ just by computation. So these two form two solutions. Now we want to use Theorem 1 (IV.1.4), so we must find when the Wronskian is non-zero.

$$W[\phi_1, \phi_2](t) = \det \begin{bmatrix} \phi_1 & \phi_2 \\ \phi_1' & \phi_2' \end{bmatrix} = \det \begin{bmatrix} te^{rt} & e^{rt} \\ e^{rt} + rte^{rt} & re^{rt} \end{bmatrix} = tre^{2rt} - e^{2rt} - rte^{2rt} = -e^{2rt},$$

which is never 0. So by Theorem 1 (IV.1.4), all solutions are of the form

$$y(t) = C\phi_1(t) + D\phi_2(t),$$

for constants $C$ and $D$.

To introduce more notation, when working with a homogeneous differential equation $y'' + p(t)y' + q(t)y = 0$, some people like to write $L[\phi]$ as the function on the left hand side of that equality:

$$L[\phi] = \phi'' + p\phi' + q\phi.$$  

Note that the differential equation is then stating $L[y] = 0$. Obviously not all functions $\phi$ have $L[\phi] = 0$, so it is up to the above methods to try and figure out which $\phi$ satisfy this. Right now, we only know how to do this when the differential equations involves constant coefficients. But supposing that we’re giving one $\phi_1$, how do we find a second solution $\phi_2$? Below is a method to help with this.

### § IV.1.A. Reduction of order

Returning to differential equations, recall that if we can find two “different enough” solutions to a second-order, ho-
mogeneous, differential equation

\[ P(t)y'' + Q(t)y' + R(t)y = 0, \]

then any solution will be a linear combination of those two solutions: \( Cy_1 + Dy_2 \) for constants \( C \) and \( D \). But often it can be hard to find two solutions, let alone one. The method of reduction of order can sometimes be used to help with this. In general, reduction of order is just a means of reducing the order of a differential equation by considering a different differential equation with a smaller order.

### IV.1.A • 1. Result 18

Consider the differential equation

\[ y'' + p(t)y' + q(t)y = 0. \]

Suppose \( y_1 \) is a solution to this (that isn’t just the constant 0). Thus

\[ y_2(t) = v(t) \cdot y_1(t) \]

is also a solution, given that we can find a \( v \) where

\[ y_1(t)v'' + (2y_1'(t) + p(t)y_1(t))v' = 0. \]

**Proof**: If we can find such a \( v \), then by the product rule,

\[
\begin{align*}
y_1' &= v'y_1 + vy_1' \\
y_2' &= v'y_1 + 2v'y_1 + vy_1'' = v'y_1 + 2v'y_1 - vpy_1' - vy_1.
\end{align*}
\]

Thus

\[
y_2'' + py_2' + qy_2 = (v''y_1 + 2v'y_1 - vpy_1' - vy_1) + pv_1'y_1 + pv_1y_1' + qvy_1
\]

\[ = v'y_1 + (2y_1' - v + py_1)u'
\]

\[ = 0. \]

The equation above used to find \( v \) is really just a result of plugging in \( y_2 = v_1y_1 \) to the differential equation, and finding out what \( v \) must satisfy. Actually finding a \( v \) where that happens can be done by instead solving for \( w = v' \) in the first-order, linear differential equation

\[ y_1w' + (2y_1 + py_1)w = 0, \]

and then integrating \( w \) to get \( v \).

### Section IV.2. Non-Homogeneous Equations

**§ IV.2.A. Theory behind non-homogeneous equations**

A homogeneous, second-order, differential equation is any equation of the form

\[ P(t)y'' + Q(t)y' + R(t)y = 0. \]

We “know” how to solve a second-order, homogeneous differential equation in general—at least in principle. Doing this requires finding two particular solutions that are “different enough”—in that they have a non-zero Wronskian. Finding two solutions can be incredibly difficult, but once we have them, \( \phi_1 \) and \( \phi_2 \) to

\[ P(t)y'' + Q(t)y' + R(t)y = 0, \]

then all solutions to this will be of the form

\[ y(t) = C\phi_1(t) + D\phi_2(t) \]

for various constants \( C \) and \( D \).

Now we will turn to non-homogeneous, linear, second-order, differential equations.
A non-homogeneous, linear, second-order, differential equation is an equation of the form
\[ P(t)y'' + Q(t)y' + R(t)y = S(t) \]
for functions \( P, Q, R, \) and \( S \).

Solving these is even harder in general, so we will look at a couple of methods to help. The guiding idea is the following result.

Consider the homogeneous differential equation
\[ P(t)y'' + Q(t)y' + R(t)y = 0. \]
Let \( \phi_1 \) and \( \phi_2 \) be solutions of this with non-zero Wronskian. Thus all solutions to the non-homogeneous differential equation
\[ P(t)y'' + Q(t)y' + R(t)y = S(t) \]
will be of the form
\[ y(t) = C\phi_1(t) + D\phi_2(t) + Y(t), \]
where \( Y(t) \) solves the non-homogeneous equation.

**Proof:**

Firstly, let \( y(t) \) be an arbitrary solution to the non-homogeneous, differential equation. Consider the function \( \phi = y - Y \). We have derivatives \( \phi' = y' - Y' \), and \( \phi'' = y'' - Y'' \). Hence
\[
P\phi'' + Q\phi' + R\phi = P \cdot (y'' - Y'') + Q \cdot (y' - Y') + R \cdot (y - Y)
\]
\[
= (Py'' + Qy') - (Py'' + QY' + RY)
\]
\[
= S - S = 0.
\]
In other words, \( \phi \) is a solution to the homogeneous differential equation
\[ P(t)y'' + Q(t)y' + R(t)y = 0. \]
But those solutions are of the form \( C\phi_1 + D\phi_2 \) for some constants \( C \) and \( D \). Thus for some constants \( C, D, \)
\[ y - Y = C\phi_1 + D\phi_2 \]
\[ \therefore y = C\phi_1 + D\phi_2 + Y \]
So if solving a second-order, homogeneous, differential equation requires finding two solutions—\( \phi_1 \) and \( \phi_2 \)—solving a non-homogeneous, linear, differential equation requires finding three—\( \phi_1 \) and \( \phi_2 \) to the homogeneous version, and \( Y \) to the non-homogeneous version. The next two ideas are trying to find such a \( Y \).

§ IV.2.B. Undetermined coefficients

The method of undetermined coefficients is in essence making an educated guess at a particular solution \( Y \) to an equation of the form
\[ ay'' + by' + cy = g(t) \]
for various \( g(t) \). Note that this is somewhat restricted for two reasons. Firstly, rather than second-order, linear, differential equations in general,
\[ P(t)y'' + Q(t)y' + R(t) = g(t), \]
we require that the coefficients \( P, Q, \) and \( R \) be constants. Secondly, \( g \) must be simple or nice enough to be able to guess the solution. This means products and sums of polynomials, sines, cosines, and exponentials. If \( g \) involves a quotient, or some other function, we can’t use the method of undetermined coefficients.
Consider the differential equation \( ay'' + by' + cy = g(t) \).

If \( g(t) = a_n t^n + \cdots + a_1 t^1 + a_0 \), then guess \( Y(t) = A_n t^n + \cdots + A_1 t^1 + A_0 \);

if \( g(t) = \sin(at) \) or \( g(t) = \cos(at) \), then guess \( Y(t) = A \sin(at) + B \cos(at) \);

if \( g(t) = e^{at} \), then guess \( Y(t) = Ae^{at} \).

for various constants.

If the guess ends up being a solution to the homogeneous equation \( ay'' + by' + cy = 0 \), then just multiply the guess by \( t \). And if that also is a solution, keep multiplying by \( t \) until it isn’t.

If \( g \) is a product/sum of these, then our guess should be a product/sum of the corresponding guesses. For example, if \( g(t) = t^3 \sin(at) e^{bt} \), we should guess

\[
Y(t) = (C_3 t^3 + C_2 t^2 + C_1 t + C_0) \cdot (A \sin(at) + B \cos(at)) \cdot Ge^{bt},
\]

for undetermined coefficients \( C_0, C_1, C_2, C_3, A, B \), and \( G \). Note we need different constants for each guess involved.

Noting the restrictions on the method of undetermined coefficients is where variation of parameters comes in.

### § IV.2.C. Variation of parameters

A better method, but harder to actually do, is variation of parameters. Although significantly more general, applying to differential equations of the form

\[
y'' + p(t)y' + q(t)y = g(t),
\]

the method is harder to work with, and harder to remember. The basic thing to remember is the setup 

\[
Y(t) = u_1 \phi_1 + u_2 \phi_2
\]

for solutions \( \phi_1 \) and \( \phi_2 \) to the homogeneous version, and

\[
\begin{bmatrix}
\phi_1 & \phi_2 \\
\phi'_1 & \phi'_2
\end{bmatrix}
\begin{bmatrix}
u_1' \\
u_2'
\end{bmatrix}
= \begin{bmatrix} 0 \\ g \end{bmatrix}.
\]

You can then solve for \( u_1 \) and \( u_2 \) to find \( Y \). This will generalize for higher-order differential equations, when we have the setup that

\[
Y = u_1 \phi_1 + \cdots + u_n \phi_n,
\]

and then assume

\[
W \cdot \begin{bmatrix} u'_1 \\
\vdots \\ u'_{n-1} \\
u'_n
\end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ g
\end{bmatrix},
\]

where \( W \) is the Wronskian matrix of \( \phi_1, \ldots, \phi_n \). So it’s good to memorize the first matrix equation.

Using either method, once we have \( \phi_1, \phi_2 \)—fundamental solutions to the homogeneous version of the differential equation—and a \( Y \), the general solution to the non-homogeneous equation will be

\[
C \phi_1 + D \phi_2 + Y
\]

for constants \( C \) and \( D \).

### IV.2.C.1. Result 21 (Variation of Parameters)

Consider the differential equation

\[
y'' + p(t)y' + q(t)y = g(t).
\]

Let \( \phi_1 \) and \( \phi_2 \) be two solutions to the homogeneous version

\[
y'' + p(t)y' + q(t)y = 0 \text{ with } W[\phi_1, \phi_2] \neq 0.
\]

Thus the following \( Y \) is a solution to the non-homogeneous equation

\[
Y = -\phi_1 \int \frac{\phi_2 g}{W[\phi_1, \phi_2]} \, dt + \phi_2 \int \frac{\phi_1 g}{W[\phi_1, \phi_2]} \, dt.
\]

So all solutions are of the form \( y = C \phi_1 + D \phi_2 + Y \) for constants \( C \) and \( D \).
Proof:...

We begin by assuming two things:

\[ Y(t) = u_1(t)\phi_1(t) + u_2(t)\phi_2(t) \]
\[ Y'(t) = u_1(t)\phi_1'(t) + u_2(t)\phi_2'(t), \]

for some functions \( u_1 \) and \( u_2 \). This ensures that \( u_1 \) and \( u_2 \) act similar to constants when we take the first derivative. The second derivative, however, allows for that to change:

\[ Y'' = \frac{d}{dt}(u_1\phi_1') + \frac{d}{dt}(u_2\phi_2') \]
\[ = u'_1\phi_1'' + u'_2\phi_2'' + u_2\phi_2'' + u_2\phi_2''' \]

Note that then \( Y \) is a solution to the given differential equation iff

\[ Y'' + pY' + qY = g(t) \]
\[ \iff \]
\[ u'_1\phi_1'' + u'_1\phi_1'' + u'_2\phi_2'' + p(u_1\phi_1' + u_2\phi_2') + q(u_1\phi_1 + u_2\phi_2) = g(t) \]
\[ \iff \]
\[ u'_1\phi_1' + u_1\phi_1'' + p\phi_1' + q\phi_1 + u'_2\phi_2' + u_2\phi_2'' + u_2\phi_2'' = g(t) \]
\[ \iff u'_1\phi_1' + u'_2\phi_2' = g(t). \]

Now because \( Y'(t) = u_1\phi_1' + u_2\phi_2' \), the rest of the terms we would get from the product rule must be 0:

\[ u'_1\phi_1 + u'_2\phi_2 = 0. \]

Hence we have two equations, and two unknowns—\( u'_1 \) and \( u'_2 \)—which is equivalent to the matrix equation

\[
\begin{bmatrix}
\phi_1 & \phi_2 \\
\phi_1' & \phi_2'
\end{bmatrix}
\begin{bmatrix}
u'_1 \\
u'_2
\end{bmatrix} = \begin{bmatrix}0 \\
g(t)
\end{bmatrix}. 
\]

Since the Wronskian—the determinant of the matrix on the left—is non-zero, we can invert the matrix to get the equivalent statement that

\[
\begin{bmatrix}
u'_1 \\
u'_2
\end{bmatrix} = \frac{1}{W[\phi_1, \phi_2]}
\begin{bmatrix}
\phi_2 & -\phi_2 \\
-\phi_1' & \phi_1
\end{bmatrix}
\begin{bmatrix}0 \\
g(t)
\end{bmatrix}.
\]

So any functions \( u_1 \) and \( u_2 \) satisfying this will have \( Y = u_1\phi_1 + u_2\phi_2 \) satisfy the differential equation. In particular, integrating tells us that

\[ u_1 = \int \frac{-\phi_2g}{W[\phi_1, \phi_2]} \, dt, \quad \text{and} \quad u_2 = \int \frac{\phi_1g}{W[\phi_1, \phi_2]} \, dt \]

work. Thus we have found a particular solution \( Y \), and so by Result 19 (IV.2.A・2), any solution will be of the form \( C\phi_1 + D\phi_2 + Y \) for constants \( C \) and \( D \).

So in general, the process is to assume

\[ Y = u_1\phi_1 + u_2\phi_2 \]

where

\[ u'_1\phi_1 + u'_2\phi_2 = 0 \]
\[ u'_1\phi_1' + u'_2\phi_2' = g, \]

and then to solve for \( u'_1 \) and \( u'_2 \) to get some \( u_1 \) and \( u_2 \) that work.
Section IV.3. Worked-Out Examples

IV.3.1. Example 31

Find the general solution to \( y'' + 2y' + 2y = 0 \).

**Solution:**

Note that \( e^{rt} \) solves this iff

\[
e^{rt}(r^2 + 2r + 2) = 0 \iff r^2 + 2r + 2 = 0 \iff r = -1 \pm i.
\]

Thus \( e^{(-1+i)t} \) and \( e^{(-1-i)t} \) form a fundamental set of solutions. We can write these without \( i \) by first using Euler’s formula:

\[
e^{(-1+i)t} = e^{-t}e^{it} = e^{-t}(\cos t + i\sin t),
\]

\[
e^{(-1-i)t} = e^{-t}e^{-it} = e^{-t}(\cos(-t) + i\sin(-t)) = e^{-t}\cos t - ie^{-t}\sin t.
\]

So we can write \( Ce^{(-1+i)t} + De^{(-1-i)t} \) as

\[
Ce^{-t}\cos t + iCe^{-t}\sin t + De^{-t}\cos t - iDe^{-t}\sin t = (C + D)e^{-t}\cos t + i(C - D)e^{-t}\sin t = C_1e^{-t}\cos t + C_2e^{-t}\sin t.
\]

where \( C_1 = C + D \), and \( C_2 = i(C - D) \) are two arbitrary constants.

IV.3.2. Example 32

Consider the initial value problem

\[
3u'' - u' + 2u = 0, \quad u(0) = 2, \quad u'(0) = 0.
\]

1. Find the solution \( u(t) \) of this problem.
2. For \( t > 0 \), find the first time at which \( |u(t)| = 10 \).

**Solution:**

1. The solutions to this of the form \( e^{rt} \) will need to have

\[
e^{rt}(3r^2 - r + 2) = 0 \iff 3r^2 - r + 2 = 0 \iff r = \frac{1 \pm \sqrt{23}}{6}.
\]

It’s then easy to confirm that the solution will be of the form

\[
u(t) = Ce^{t/6}\cos(\sqrt{23t}/6) + De^{t/6}\sin(\sqrt{23t}/6)
\]

for some constants \( C \) and \( D \). The requirement that \( u(0) = 2 \) has \( 2 = C \). Write \( A \) for \( \sqrt{23}/6 \) so that

\[
u(t) = Ce^{t/6}\cos(At) + De^{t/6}\sin(At)
\]

\[
u'(t) = \frac{C}{6}e^{t/6}\cos(At) - ACe^{t/6}\sin(At) + \frac{D}{6}e^{t/6}\sin(At) + ADe^{t/6}\cos(At).
\]

\[
\therefore u'(0) = \frac{C}{6} + AD = \frac{2}{6} + \frac{\sqrt{23}}{6}D
\]

So the initial condition \( u'(0) = 0 \) requires that \( D = -\frac{2}{\sqrt{23}} \). Hence

\[
u(t) = 2e^{t/6}\cos(\sqrt{23t}/6) - \frac{2}{\sqrt{23}}e^{t/6}\sin(\sqrt{23t}/6).
\]

2. Using a plotter, we can see the first time at which \( |u(t)| = 10 \) is between 9 and 12, and has \( u(t) = -10 \):
Consider the initial value problem

\[ 5u'' + 2u' + 7u = 0, \quad u(0) = 2, \quad u'(0) = 1. \]

1. Find the solution \( u(t) \) of this problem.
2. Find the smallest \( T \) such that \( |u(t)| \leq 0.1 \) for all \( t > T \).

**Solution:**

1. Solving for \( r \) in

\[ 5r^2 + 2r + 7 = 0 \]

so that for some constants \( C \) and \( D \),

\[ u(t) = Ce^{-t/5} \cos(\sqrt{34}t/5) + De^{-t/5} \sin(\sqrt{34}t/5). \]

Writing \( A \) for \( \sqrt{34}/5 \), we have

\[ u(t) = Ce^{-t/5} \cos(At) + De^{-t/5} \sin(At) \]

so \( u(0) = C \)

\[ u'(t) = -\frac{C}{5}e^{-t/5} \cos(At) - ACe^{-t/5} \sin(At) - \frac{D}{5}e^{-t/5} \sin(At) + ADe^{-t/5} \cos(At). \]

so \( u'(0) = -\frac{C}{5} + AD \).

Hence the initial condition \( u(0) = 2 \) implies \( C = 2 \), and the initial condition \( u'(0) = 1 \) implies \( AD = 7/5 \), or that \( D = 7/\sqrt{34} \). Hence

\[ u(t) = 2e^{-t/5} \cos(\sqrt{34}t/5) + \frac{7}{\sqrt{34}}e^{-t/5} \sin(\sqrt{34}t/5). \]

2. Again, we must use a calculator to find the first \( t \) where this happens. Note that for \( y(t) = \frac{17}{5}e^{-t/5} \),

\[ -y(t) \leq u(t) \leq y(t). \]

So if \( y(t) < .1 \), then \( |u(t)| < .1 \). Note that \( y(t) < .1 \) for \( t > 5 \ln(34) \approx 17.6318 \), so we don’t need to confirm \( |u(t)| < .1 \) for \( t > 18 \). Using the following plot,

We can see the last time \( u(t) \) is .1 is \( T \) found somewhere between 14 and 15. Using a calculator, we can say that this value is \( T \approx 14.5115 \).
IV.3 • 4. Example 34

Show that \( W[e^{\lambda t}\cos(\mu t), e^{\lambda t}\sin(\mu t)] = \mu e^{2\lambda t} \).

Solution:

By definition of the Wronskian, this is \( fg' - f'g \) for \( f(t) = e^{\lambda t}\cos(\mu t) \) and \( g(t) = e^{\lambda t}\sin(\mu t) \). We have the following table calculating derivatives.

\[
\begin{align*}
  f(t) &= e^{\lambda t}\cos(\mu t) \\
  g(t) &= e^{\lambda t}\sin(\mu t) \\
  f'(t) &= \lambda e^{\lambda t}\cos(\mu t) - \mu e^{\lambda t}\sin(\mu t) \\
  g'(t) &= \lambda e^{\lambda t}\sin(\mu t) + \mu e^{\lambda t}\cos(\mu t)
\end{align*}
\]

Thus we can calculate

\[
W(t) = \mu e^{2\lambda t}\cos(\mu t) - \mu e^{2\lambda t}\sin(\mu t) + \lambda e^{2\lambda t}\sin(\mu t) + \mu e^{2\lambda t}\cos(\mu t) = \mu e^{2\lambda t}.
\]

IV.3 • 5. Example 35

An equation of the form

\[
t^2\frac{d^2y}{dt^2} + \alpha t \frac{dy}{dt} + \beta y = 0, \quad t > 0,
\]

where \( \alpha \) and \( \beta \) are real constants, is called an Euler equation.

a. Let \( x = \ln t \) and calculate \( \frac{dy}{dx} \) and \( \frac{d^2y}{dx^2} \) in terms of \( \frac{dy}{dt} \) and \( \frac{d^2y}{dt^2} \).

b. Use the results of part (a) to transform equation (33) into

\[
\frac{d^2y}{dx^2} + (\alpha - 1) \frac{dy}{dx} + \beta y = 0.
\]

Observe that differential equation (34) has constant coefficients. If \( y_1(x) \) and \( y_2(x) \) form a fundamental set of solutions of equation (34), then \( y_1(\ln t) \) and \( y_2(\ln t) \) form a fundamental set of solutions of equation (33).

Solution:

a. Since we’re assuming \( t > 0 \), we can write \( \ln t = x \). Using the chain rule,

\[
\frac{dy}{dt} = \frac{dy}{dx} \frac{dx}{dt} = \frac{dy}{dx} \frac{1}{t}.
\]

Using the product rule on this to take the second derivative, we get

\[
\frac{d^2y}{dt^2} = \frac{d^2y}{dx^2} \frac{dx}{dt} \frac{dt}{dx} + \frac{dy}{dx} \frac{d}{dt} \left( \frac{1}{t} \right) = \frac{d^2y}{dx^2} \frac{1}{t^2} + \frac{dy}{dx} \frac{1}{t^2} = \frac{d^2y}{dx^2} \frac{1}{t^2} + \frac{dy}{dx} \frac{1}{t^2}.
\]

b. Note that \( t = e^x \) so that substituting our equalities so far into (33) yields that

\[
0 = e^{2x} \left( \frac{d^2y}{dx^2} \frac{1}{e^{2x}} - \frac{dy}{dx} \frac{1}{e^{2x}} \right) + \alpha e^{x} \frac{dy}{dx} \frac{1}{e^{2x}} + \beta y
\]

\[
= \left( \frac{d^2y}{dx^2} - \frac{dy}{dx} \right) + \alpha \frac{dy}{dx} + \beta y
\]

\[
= \frac{d^2y}{dx^2} + (\alpha - 1) \frac{dy}{dx} + \beta y.
\]
IV.3 • 6. Example 36

Find the general solution of $y'' - 2y' + y = 0$.

Solution ::

Solutions to $r^2 - 2r + 1 = 0$ are just $r = 1$. Thus the general solution by Result 17 (IV.1 • 7) is

$$y(t) = Ce^t + Dte^t$$

for constants $C$ and $D$.

IV.3 • 7. Example 37

Find the general solution of $16y'' + 24y' + 9y = 0$.

Solution ::

Solutions to $16r^2 + 24r + 9 = 0$ are just $r = -16/18 = -3/4$. Thus the general solution by Result 17 (IV.1 • 7) is

$$y(t) = Ce^{-3t/4} + Dte^{-3t/4}$$

for constants $C$ and $D$.

IV.3 • 8. Example 38

Solve the initial value problem

$$y'' + 4y' + 4y = 0, \quad y(-1) = 2, \quad y'(-1) = 1.$$  

Sketch the graph of the solution and describe its behavior for increasing $t$.

Solution ::

The solutions to $r^2 + 4r + 4 = 0$ are just $r = -2$ so that previous work implies

$$y(t) = Ce^{-2t} + Dte^{-2t}$$

$$\therefore \ y'(t) = -2Ce^{-2t} + De^{-2t} - 2Dte^{-2t}.$$  

So the requirement that $y(-1) = 2$ says that

$$2 = Ce^2 - De^2 \iff C - D = 2e^{-2} \iff C = 2e^{-2} + D.$$  

And the requirement that $y'(-1) = 1$ says that

$$1 = -2Ce^2 + De^2 + 2Dte^2 = e^2(-2C + 3D).$$  

The two together yield that $C = 2e^{-2} + D$ so that

$$e^{-2} = -2(2e^{-2} + D) + 3D = -4e^{-2} + D \iff D = 5e^{-2} \implies C = 7e^{-2}.$$  

Thus we can write

$$y(t) = 7e^{-2t} + 5te^{-2t}.$$  

So as $t \to \infty$, since $\lim_{t \to \infty} te^{-2t} = 0$ by l’Hôpital’s rule, $y(t) \to 0$.  

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Solve the equation \( y'' + y' + 3y = 0 \) with initial conditions \( y(0) = 1 \) and \( y'(0) = 1 \).

**Solution:**

Solutions of the form \( e^{rt} \) will require
\[
e^{rt}(r^2 + r + 3) = 0 \quad \iff \quad r = -\frac{1 \pm i \sqrt{11}}{2}.
\]
Hence the general solution will be
\[
y(t) = Ce^{-t/2} \sin(\sqrt{11}t/2) + De^{-t/2} \cos(\sqrt{11}t/2)
\]
\[
\therefore \quad y'(t) = \frac{-y(t)}{2} + \frac{\sqrt{11}}{2}Ce^{-t/2} \cos(\sqrt{11}t/2) - \frac{\sqrt{11}}{2}De^{-t/2} \sin(\sqrt{11}t/2).
\]
The initial conditions yield that
\[
1 = D
\]
\[
1 = -\frac{1}{2} + \frac{\sqrt{11}}{2}C \quad \rightarrow \quad C = \frac{3}{\sqrt{11}}
\]
Hence the final answer is
\[
y(t) = \frac{3}{\sqrt{11}}e^{-t/2} \sin(\sqrt{11}t/2) + e^{-t/2} \cos(\sqrt{11}t/2).
\]

**Example 40**

a. Consider the equation \( y'' + 2ay' + a^2 y = 0 \). Show that the roots of the characteristic equation are \( r_1 = r_2 = -a \) so that one solution of the equation is \( e^{-at} \).

b. Use Abel’s theorem \([W(t)] = ce^{\int \mu(t) \, dt} \) for \( c \) a constant, and \( p(t) \) in \( y'' + p(t)y' + q(t)y = 0 \) to show that the Wronskian of any two solutions of the given equation is
\[
W(t) = (y_1(t)y_2'(t) - y_1'(t)y_2(t)) = ce^{-2at},
\]
where \( c \) is a constant.

c. Let \( y_2(t) = e^{-at} \) and use the result of part (b) to obtain a differential equation satisfied by a second solution \( y_2(t) \). By solving this equation, show that \( y_2(t) = te^{-at} \).

**Solution:**

1. Note that \((r + a)^2 = r^2 + 2a + a^2\), which is the characteristic equation. This is 0 if \( r + a = 0 \), i.e., if \( r = -a \). So \( -a \) is the only solution. But as a solution, \( e^{-at} \) is a solution to the differential equation.

2. Abel’s formula says that the Wronskian of any two solutions will be \( ce^{\int \mu(t) \, dt} = Ce^{-2at} \) for some constant \( C \).

3. The definition of the Wronskian is \( y_1y_2' - y_1'y_2 \) for solutions \( y_1 \) and \( y_2 \). Given that \( e^{-at} = y_1(t) \) works, we would like to solve the differential equation for \( y_2 \):
\[
Ce^{-2at} = e^{-at}y_2 + ae^{-at} \quad \iff \quad Ce^{-2at} = y_2 + ay_2.
\]

Using an integrating factor of \( \mu(t) = e^{at} \) yields that \( C = (\mu y_2)' \) and thus
\[
y_2(t) = \frac{1}{\mu(t)} \frac{Ct + D}{\mu(t)} = Cte^{-at} + De^{-at},
\]
for some constant \( D \). This means any constants \( C \neq 0 \) and \( D \) will work. So one such solution is when \( C = 1 \) and \( D = 0 \), i.e.
\[
y_2(t) = te^{-at}.
\]
IV.3.11. Example 41

Use the method of reduction of order to find a second solution of the differential equation
\[ t^2 y'' + 2ty' - 2y = 0, \quad t > 0; \quad y_1(t) = t. \]

**Solution:**

The differential equation in a more usable form is, since we’re assuming \( t > 0 \),
\[ y'' + \frac{2}{t} y' - \frac{2}{t^2} y = 0. \]

The method of reduction of order says that we want a \( v \) where
\[ y_1(t)v'' + (2y_1'(t) + p(t)y_1(t))v' = 0 \]
so that \( y_2 = v \cdot y_1 \) is another solution. In our case, \( y_1(t) = t \) so that \( y_1'(t) = 1 \), and \( p(t) = 2/t \). Thus we want a \( v \) where
\[ 0 = tv'' + \left(2 + \frac{2}{t}\right)v' - \frac{2}{t^2} v = tv'' + 4v'. \]
Since \( t > 0 \), we can divide by it, and find an integrating factor of \( t^4 \) to get
\[ 0 = t^4v'' + 4t^3v' \quad \overset{\text{or}}{\Longrightarrow} \quad (t^4v')' = 0 \quad \overset{\text{or}}{\Longrightarrow} \quad v' = \frac{C}{t^4} \quad \overset{\text{or}}{\Longrightarrow} \quad v(t) = \frac{C}{t^3} + D. \]
Hence \( C = 1 \) and \( D = 0 \) work so that \( v(t) = \frac{1}{t^3} \) is a solution to our requirement. Thus by the method of reduction, we get another solution \( y_2(t) \) given by
\[ y_2(t) = v(t)y_1(t) = \frac{1}{t^3}t = \frac{1}{t^2}. \]

IV.3.12. Example 42

The differential equation
\[ y'' + \delta(xy' + y) = 0 \]
arises in the study of turbulent flow of a uniform stream past a circular cylinder.

1. Verify that \( y_1(x) = \exp(-\delta x^2/2) \) is one solution, and then
2. find the general solution in the form of an integral.

**Solution:**

1. To verify that this is a solution, we can calculate the derivatives by
   \[ y_1'(x) = -\delta x \exp(-\delta x^2/2) = -\delta xy_1(x) \]
   \[ y_1''(x) = -\delta y_1(x) + \delta^2 x^2 y_1(x) \]

   Hence the left-hand-side of the differential equation becomes
   \[ y_1''(x) + \delta(xy_1'(x) + y(x)) = -\delta y_1(x) + \delta^2 x^2 y_1(x) + \delta(-\delta x^2 y_1(x) + y_1(x)) \]
   \[ = -\delta y_1(x) + \delta^2 x^2 y_1(x) - \delta^2 x^2 y_1(x) + \delta y_1(x) \]
   \[ = -\delta y_1(x) + \delta y_1(x) \]
   \[ = 0. \]

2. To find another solution, proceed by reduction of order. Recall that for
   \[ y'' + p(t)y' + q(t)y = 0, \]
   and solution \( y_1(t) \)—that isn’t constantly \( 0 \)—if there is a function \( v \) where
   \[ y_1v'' + (2y_1' + p y_1)v' = 0, \]
then we can get a second solution \( y_2 = v \cdot y_1 \).
In our case,

\[ p(x) = \delta x \]
\[ q(x) = \delta, \]

using \( y_1 \) as the given solution— which is never 0—we would like a \( v(x) \) that satisfies

\[ 0 = y_1 v'' + (2y_1 + \delta xy_1) v' \]
\[ = y_1 v'' + (-2\delta xy_1 + \delta xy_1) v' \]
\[ = y_1 (v'' - \delta xv') \]
\[ \longleftrightarrow 0 = v'' - \delta xv'. \]

So we want an integrating factor \( \mu(x) \) satisfying \( \mu' = -\delta x \mu \), e.g. \( \mu = e^{-\delta x^2/2} \). In this case, the above is equivalent to

\[ (\mu v')' = 0 \longleftrightarrow v' = \frac{c}{\mu} = ce^{\delta x^2/2}, \]

for some constant \( c \). As any constant \( c \) will work, \( c = 1 \) also works. Hence \( v(x) = \int e^{\delta x^2/2} \, dx \) is a sufficient function. Thus by the method of reduction of order,

\[ y_2(x) = v(x) \cdot y_1(x) = e^{-\delta x^2/2} \int e^{\delta x^2/2} \, dx \]

is another solution. Thus the general solution to the differential equation \( y'' + \delta (xy' + y) = 0 \) is

\[ Ce^{-\delta x^2/2} + De^{-\delta x^2/2} \int e^{\delta x^2/2} \, dx \]

for constants \( C \) and \( D \).

----

**IV.3 • 13. Example 43**

Consider the differential equation

\[ y'' + y = 3 + 4\sin(2t). \]

Give the general solution.

**Solution . . .**

Since the non-homogeneous part is the sum of a polynomial—of degree 0—and a sine function, our guess for a \( Y \) should be of the form \( A + B\sin(2t) + C\cos(2t) \) for constants \( A, B, \) and \( C \). We only need one solution, so we don’t need all the different constants that work. First let’s calculate the derivatives:

\[ Y(t) = A + B\sin(2t) + C\cos(2t) \]
\[ Y'(t) = 2B\cos(2t) - 2C\sin(2t) \]
\[ Y''(t) = -4B\sin(2t) - 4C\cos(2t). \]

\[ Y'' + Y = A + \sin(2t)(-4B + B) + \cos(2t)(-4C + C) \]
\[ = A - 3B\sin(2t) - 3C\cos(2t). \]

So this equals \( 3 + 4\sin(2t) \) iff the coefficients match up: \( A = 3 \) and \( -3B = 4 \) with \( C = 0 \). In other words, we get

\[ Y(t) = 3 - \frac{3}{4}\sin(2t). \]

Now we want to solve the homogeneous differential equation

\[ y'' + y = 0. \]

To do this, we proceed as before: consider solutions of the form \( e^{rt} \), which requires

\[ r^2 + 1 = 0 \longleftrightarrow r = \pm i. \]

Thus our solutions are of the form \( C_1 e^{it} + C_2 e^{-it} = C \cos(t) + D \sin(t) \) for various constants. Therefore, by
Result 19 (IV.2.A • 2), the general solution to the non-homogeneous, differential equation is
\[ y(t) = C\cos(t) + D\sin(t) + 3 - \frac{3}{4}\sin(2t) \]
for various constants \(C\) and \(D\).

### IV.3.14. Example 44

Consider the differential equation
\[ y'' + y = \tan t. \]
Find the general solution.

**Solution:**

First let’s solve the homogeneous, differential equation
\[ y'' + y = 0, \]
which has solutions \(e^{it}\) and \(e^{-it}\). Thus the general form for the solutions of this homogeneous differential equation is
\[ \phi(t) = C_1e^{it} + C_2e^{-it} = C\cos t + D\sin t \]
for various constants. In other words, \(\phi_1(t) = \cos t\) and \(\phi_2(t) = \sin t\) are two such solutions. Now their Wronskian is \(W[\cos t, \sin t] = \cos^2 t + \sin^2 t = 1\), so by Variation of Parameters (IV.2.C • 1), skipping the steps in integration,

\[
\begin{align*}
    u_1 & := \int \frac{-\sin t \tan t}{W[\cos t, \sin t]} \, dt \\
    & = \int \frac{-\sin^2 t}{\cos t} \, dt \\
    & = \sin t - \ln |\sec t + \tan t| \\
    u_2 & := \int \frac{\cos t \tan t}{W[\cos t, \sin t]} \, dt \\
    & = \int \sin t \, dt \\
    & = -\cos t.
\end{align*}
\]

the constants of integration are left out, since they don’t change whether \(u_1\) and \(u_2\) work. Thus \(Y = u_1\phi_1 + u_2\phi_2\) is a solution to the non-homogeneous, differential equation, and so the general solution is
\[ y(t) = C\phi_1(t) + D\phi_2(t) + u_1(t)\phi_1(t) + u_2(t)\phi_2(t) \]
for constants \(C\) and \(D\).

### IV.3.15. Example 45

Find the general solution to \(y'' - 2y' - 3y = -3te^{-t}\).

**Solution:**

First we find the solutions to the homogeneous version:
\[ \phi'' - 2\phi' - 3\phi = 0 \]
yields characteristic equation
\[ r^2 - 2r - 3 = 0 \quad \leftrightarrow \quad r = -1 \text{ or } r = 3. \]
Hence \(\phi_1(t) = e^{-t}\) and \(\phi_2(t) = e^{3t}\) yield the general solution to the homogeneous version: \(C\phi_1 + D\phi_2\).

Now we need a particular solution to the non-homogeneous equation. Using the method of undetermined
coefficients, our guess should be the product of a polynomial of degree 1, and \(e^{-t}\):

\[ Y(t) = (At + B) \cdot e^{-t}. \]

No constant is needed in front of the \(e^{-t}\), since we can just distribute that into the constants of \(At + B\). We can calculate

\[ Y' = Ae^{-t} - Y \]
\[ Y'' = -2Ae^{-t} + Y. \]

This solves the non-homogeneous equation iff

\[ Y'' - 2Y' - 3Y = -3te^{-t} \iff -3te^{-t} = [ -2Ae^{-t} + Y ] - 2[Ae^{-t} - Y] - 3Y \]
\[ \iff -3te^{-t} = -2Ae^{-t} + Y - 2Ae^{-t} + 2Y - 3Y \]
\[ \iff -3te^{-t} = -4Ae^{-t}, \]

which requires \(A = 0\), which would require \(Y\) to be a solution to the homogeneous equation. In other words, this won’t work. So let’s multiply our guess by \(t\) to see whether that does the trick.

\[ Y = (At^2 + Bt)e^{-t} \]
\[ Y' = (2At + B)e^{-t} - Y \]
\[ Y'' = 2Ae^{-t} - (2At + B)e^{-t} - Y' \]
\[ = 2Ae^{-t} - 2(2At + B)e^{-t} + Y. \]

This solves the non-homogeneous equation iff

\[ Y'' - 2Y' - 3Y = -3te^{-t} \iff -3te^{-t} = [2Ae^{-t} - 2(2At + B)e^{-t} + Y] - 2[(2At + B)e^{-t} - Y] - 3Y \]
\[ \iff -3te^{-t} = 2Ae^{-t} - 2(2At + B)e^{-t} - 2(2At + B)e^{-t} \]
\[ \iff -3t = 2A - 4At - 2B - 4At - 2B \]
\[ \iff -3t = (2A - 4B) - 8At, \]

which requires \(-8A = -3\)—i.e. \(A = 3/8\) and \(2A - 4B = 0\)—i.e. \(B = A/2\), so that \(B = 3/16\). Hence we arrive at a particular solution

\[ Y = \left(\frac{3}{8}t^2 + \frac{3}{16}t\right)e^{-t}, \]

meaning the general solution to the non-homogeneous equation is

\[ y(t) = Ce^{-t} + De^{3t} + \left(\frac{3}{8}t^2 + \frac{3}{16}t\right)e^{-t}. \]

---

**IV.3 • 16. Example 46**

Find the general solution to \(y'' + y = 3 \sin(2t) + t \cos(2t)\).

**Solution:**

Firstly, the general solution to the homogeneous version will be \(C \cos t + D \sin t\). To find a particular solution to the non-homogeneous version, guess the sum of guesses

\[ Y(t) = A \sin(2t) + (Bt + C) \cos(2t) \]
\[ Y'(t) = (2A + B) \cos(2t) - 2(Bt + C) \sin(2t) \]
\[ Y''(t) = -2(2A + B) \sin(2t) - 2B \sin(2t) - 4(Bt + C) \cos(2t) \]
\[ = -4(A + B) \sin(2t) - 4(Bt + C) \cos(2t) \]

This solves the non-homogeneous equation iff

\[ 3 \sin(2t) + t \cos(2t) = Y'' + Y \]
\[ = -4(A + B) \sin(2t) - 4(Bt + C) \cos(2t) + A \sin(2t) + (Bt + C) \cos(2t) \]
\[ (-3A - 4B) \sin(2t) - 3(Bt + C) \cos(2t). \]

This requires \(-3(Bt + C) = t\)—i.e. \(C = 0\) and \(B = -1/3\)—as well as \(-3A - 4B = 3\). Given that we know \(B\), this implies

\[-3A + \frac{4}{3} = 3 \iff -3A = 5/3 \iff A = -5/9.\]

Hence our particular solution is

\[ Y = -\frac{5}{9} \sin(2t) - \frac{1}{3} \cos(2t), \]

and so the general solution to the non-homogeneous equation is

\[ y(t) = C \cos t + D \sin t - \frac{5}{9} \sin(2t) - \frac{1}{3} \cos(2t). \]

---

**IV.3 • 17. Example 47**

Find the general solution to

\[ y'' + y' + 4y = 2 \sinh t \]

given that \(\sinh t = (e^t - e^{-t})/2\).

**Solution:**

With what we know so far, it’s easy to show the general solution to the homogeneous version is

\[ Ce^{t/2} \cos(\sqrt{15}/2) + De^{t/2} \sin(\sqrt{15}/2), \]

so now our task is to find a particular solution. Given that \(g(t)\) is the sum of \(e^t\) and \(-e^{-t}\), our guess is

\[ Y = Ae^t + Be^{-t} = y' \]
\[ Y' = Ae^t - Be^{-t}. \]

So this solves the differential equation iff

\[ 2 \sinh t = e^t - e^{-t} = Ae^t + Be^{-t} + Ae^t - Be^{-t} + 4Ae^t + 4Be^{-t} \]
\[ = 6Ae^t + 4Be^{-t}, \]

which happens iff \(A = 1/6, \text{and} \ B = -1/4\). Thus our particular solution is \(Y(t) = e^t/6 - e^{-t}/4\), and the general solution is

\[ y(t) = Ce^{t/2} \cos \left( \frac{\sqrt{15}}{2} t \right) + De^{t/2} \sin \left( \frac{\sqrt{15}}{2} t \right) + \frac{1}{6} e^t - \frac{1}{4} e^{-t}. \]

---

**IV.3 • 18. Example 48**

Solve the equation

\[ y'' + 2y' + 5y = t + 2 \sin t, \]

with initial conditions \(y(0) = 1\), and \(y'(0) = 0\).

**Solution:**

First let’s find a particular solution to the equation. To do this, proceed by undetermined coefficients. Our guess should be the sum of a first-degree polynomial and sinusoidal functions:

\[ Y(t) = At + B + C \sin t + D \cos t \]
\[ Y'(t) = A + C \cos t - D \sin t \]
\[ Y''(t) = -C \sin t - D \cos t. \]

This solves the equation iff

\[ t + 2 \sin t = -C \sin t - D \cos t + 2A + 2C \cos t - 2D \sin t + 5At + 5B + 5C \sin t + 5D \cos t. \]
Chapter IV – Section 3

\[ = 5At + (2A + 5B) + (4C - 2D) \sin t + (7C - D) \cos t. \]

By equating coefficients, this tells us that \(5A = 1,\) and \(2A + 5B = 0,\) meaning \(A = 1/5,\) and \(B = -2/25.\) It also tells us that \(4C - 2D = 2\) while \(7C - D = 0,\) meaning \(C = -1/5\) and \(D = -7/5.\) Therefore our particular solution is

\[ Y(t) = \frac{1}{5}t - \frac{2}{25} - \frac{1}{5} \sin t - \frac{7}{5} \cos t. \]

Now we must find the general solution to the homogeneous version: \(y'' + 2y' + 5y = 0.\) This has solutions of the form \(e^r\) for \(r\) satisfying \(r^2 + 2r + 5 = 0,\) i.e. \(r = -1 \pm 2i.\) This yields the general solution for the homogeneous version as

\[ Ce^{-t} \sin(2t) + De^{-t} \cos(2t). \]

Therefore the general solution to the non-homogeneous equation is

\[ y(t) = Ce^{-t} \sin(2t) + De^{-t} \cos(2t) + \frac{1}{5}t - \frac{2}{25} - \frac{1}{5} \sin t - \frac{7}{5} \cos t. \]

Using the initial values, this requires that

\[ 1 = D - \frac{2}{25} - \frac{7}{5}, \]
\[ 0 = 2C - D, \]

so that \(D = 62/25,\) and thus \(C = 62/50.\) Therefore the solution to the initial value problem is

\[ y(t) = \frac{62}{50}e^{-t} \sin(2t) + \frac{62}{25}e^{-t} \cos(2t) + \frac{1}{5}t - \frac{2}{25} - \frac{1}{5} \sin t - \frac{7}{5} \cos t. \]

Consider the equation \(y'' - 3y' - 4y = 2e^{-t}\) with

\[ \phi_1(t) = e^{-t}, \quad \text{and} \quad \phi_2(t) = e^{4t}, \]

solutions to the homogeneous version. Seek a solution to the non-homogeneous equation of the form

\[ Y(t) = v(t) \cdot \phi_1(t) = v(t)e^{-t} \]

for some \(v.\)

a. Substitute \(Y, Y',\) and \(Y''\) into the non-homogeneous equation and show that \(v\) must satisfy \(v'' - 5v' = 2.\)

b. Let \(w = v'\) and show that \(w\) satisfies \(w' - 5w = 2,\) solving this for \(w.\)

c. Integrate \(w\) to find \(v\) and then show that

\[ Y = -\frac{2}{5}te^{-t} + \frac{1}{5}e^{4t} + c_1e^{-t} + c_2e^{-t}. \]

The first term on the right-hand side is the desired particular solution of the nonhomogeneous equation. Note that it is a product of \(t\) and \(e^{-t}.\)

**Solution:**

a. First calculate the derivatives.

\[ Y = ve^{-t}, \]
\[ Y' = v'e^{-t} - Y, \]
\[ = e^{-t}(v' - v) \]
\[ Y'' = v''e^{-t} - v'e^{-t} - Y' \]
\[ = e^{-t}(v'' - 2v' + v). \]

As a result, this is a solution to the non-homogeneous equation only if

\[ Y'' - 3Y' - 4Y = 2e^{-t} \iff e^{-t}(v'' - 2v' + v) - 3e^{-t}(v' - v) - 4e^{-t}v = 2e^{-t} \]
\[ \iff v'' - 2v' + v - 3(v' - v) - 4v = 2 \]

54
Solution

If there is only one solution,

If there are two imaginary solutions,

If there are two real solutions,

Let \( Y \)

Consider the characteristic equation

we know how to solve such equations.

for some constant \( C \).

c. Integrating \( w \) clearly yields

for some constant \( D \). Hence we get, as desired,

\[
Y = \psi \phi_1 = \frac{1}{5} C e^{4t} - \frac{2}{5} t e^{-t} + D e^{-t}.
\]

Consider \( ay'' + by' + cy = g(t) \) where \( a, b, c \) are positive.

Let \( Y_1 \) and \( Y_2 \) be solutions to this. Show that \( Y_1 - Y_2 \rightarrow 0 \) as \( t \rightarrow \infty \). Is this true if \( b = 0 \)?

**Solution:**

Note that the difference of any two solutions to this non-homogeneous, linear, second-order, ordinary, differential equation will be solutions to the homogeneous version, that is \( Y_1 - Y_2 = \phi \) has \( a\phi'' + b\phi' + c\phi = 0 \). But we know how to solve such equations.

Consider the characteristic equation \( ar^2 + br + c = 0 \).

1. If there is only one solution, \( r \), then this expression can be factored:

\[
ax^2 + bx + c = a(x - r)(x - r) = ax^2 - 2arx + ar^2.
\]

In other words, \( b = -2ar \), and \( c = ar^2 \). The requirement that all the coefficients be positive requires \( a > 0 \) and \(-2ar > 0\), i.e. \( r < 0 \). The solutions to the homogeneous differential equation are then

\[
\phi(t) = Ce^{rt} + Dte^{rt},
\]

which all go to 0 as \( t \rightarrow \infty \), regardless of \( C \) and \( D \), because \( r < 0 \).

2. If there are two real solutions, \( r_1 \) and \( r_2 \), then we can factor

\[
ax^2 + bx + c = a(x - r_1)(x - r_2) = ax^2 - a(r_1 + r_2)x + ar_1r_2.
\]

Again, the requirement that all the coefficients are positive puts two restrictions on \( r_1 \) and \( r_2 \). Since \( a > 0 \) and \( ar_1r_2 > 0 \), \( r_1 \) and \( r_2 \) must share the same sign: both positive or both negative. If both are positive, then \( b = -r_1 - r_2 \) would be negative. So the assumption that \( b > 0 \), tells us that \( r_1 \) and \( r_2 \) are negative. But then

\[
\phi(t) = Ce^{r_1 t} + D e^{r_2 t}
\]

goes to 0 as \( t \rightarrow \infty \), regardless of \( C \) and \( D \).

3. If there are two imaginary solutions, \( A \pm Bt \), then we can factor

\[
ax^2 + bx + c = a(x - A - Bt)(x - A + Bt) = ax^2 - 2ax + a(A^2 + B^2).
\]

The requirement that \(-2aA > 0\) tells us that \(-2A > 0\), i.e. \( A < 0 \). Since \( |\cos \theta| \) and \( |\sin \theta| \) are both at
most 1,
\[ |\phi(t)| = |Ce^{At}\cos(Bt) + De^{At}\sin(Bt)| \leq |Ce^{At}| + |De^{At}|, \]
which goes to 0 as \( t \to \infty \), since \( A < 0 \). Hence \( \phi(t) \to 0 \).

Thus in each case \( Y_1 - Y_2 = \phi(t) \) goes to 0 as \( t \to \infty \), showing what we wanted.

If \( b = 0 \), this isn’t necessarily true. For instance, \( y'' + y = 1 \) (so setting \( a = 1, b = 0, c = 1, \) and \( g = 1 \)) has solutions
\[ Y_1 = 1 + 2\cos t, \quad \text{and} \quad Y_2 = 1 + \cos t, \]
but \( Y_1 - Y_2 = \cos t \) does not go to 0 as \( t \to \infty \).

**IV.3 • 21. Example 51**

Consider the equation \( y'' + by' + cy = g(t) \), and suppose \( r_1 \) and \( r_2 \) (not necessarily distinct nor real) are zeroes of the characteristic polynomial of the corresponding homogeneous equation.

1. Verify that the non-homogeneous differential equation can be written as
\[ (D - r_1)(D - r_2)y = g(t) \]
where \( r_1 + r_2 = -b \) and \( r_1r_2 = c \).

2. Let \( u = (D - r_2)y \). Solve \( y'' + by' + cy = g(t) \) by solving the two equations
\[ (D - r_1)u = g(t), \quad (D - r_2)y = u(t). \]

**Solution . . .**

1. Firstly, we can confirm that \( (D - r_2)y = y' - r_2y \). Now applying \( D - r_1 \) to this yields
\[ (D - r_1)(y' - r_2y) = D(y' - r_2y) - r_1y' + r_1r_2y = y'' - (r_2 + r_1)y' + r_1r_2y. \]

By assumption on \( r_1 \) and \( r_2 \), this is just \( y'' - by + cy \), and hence the equation \( y'' + by' + cy = g(t) \) is equivalent to \( (D - r_1)(D - r_2)y = g(t) \).

2. Firstly, we get \( u' - r_1u = g(t) \) implies that
\[ e^{-r_1t}u = \int e^{-r_1t}g(t) \, dt + C \quad \Rightarrow \quad u = e^{r_1t} \int e^{-r_1t}g(t) \, dt + Ce^{r_1t}. \]

Using this, to find a \( y \) where \( (D - r_2)y = y' - r_2y = u \), we can again proceed as with any first-order, linear, differential equation:
\[ e^{-r_2t}y = \int e^{-r_2t}u \, dt + D \quad \Rightarrow \quad y = e^{r_2t} \int e^{-r_2t}u \, dt + De^{r_2t}. \]

Re-writing this yields
\[ y(t) = e^{r_2t} \int e^{-r_2t}u \, dt + De^{r_2t} \]
\[ = e^{r_2t} \left[ \int e^{-r_1r_2}\int e^{-r_1z}g(z) \, dz + Ce^{r_1r_2} \right] \, dt + De^{r_2t} \]
\[ = e^{r_2t} \left[ \int e^{-r_1r_2}\int e^{-r_1z}g(z) \, dz \, dt + Ce^{r_1r_2} + De^{r_2t} + c, \right. \]
for some constants \( c, C, \) and \( D \).

**IV.3 • 22. Example 52**

Use the method of variation of parameters to find a particular solution to
\[ y'' - 5y' + 6y = 2e^t \]
Solution ..

Firstly, let’s solve the homogeneous version to get
\[ r^2 - 5r + 6 = 0 \quad \iff \quad r = -2 \text{ or } r = -3. \]

Hence the general solution to the homogeneous version is \( Ce^{-2t} + De^{-3t} \). So let \( \phi_1(t) = e^{-2t} \) and \( \phi_2(t) = e^{-3t} \).

Using this, variation of parameters tells us we have a \( Y = u_1\phi_1 + u_2\phi_2 \) where \( u_1 \) and \( u_2 \) satisfy
\[
\begin{bmatrix}
\phi_1 & \phi_2 \\
\phi_1' & \phi_2'
\end{bmatrix}
\begin{bmatrix}
u_1' \\
u_2'
\end{bmatrix}
= \begin{bmatrix} 0 \\ g \end{bmatrix}.
\]

Now \( g \) here is just \( 2e^t \). We can calculate \( \phi_1' \) and \( \phi_2' \) to get the matrix equation
\[
\begin{bmatrix}
e^{-2t} & e^{-3t} \\
-2e^{-2t} & -3e^{-3t}
\end{bmatrix}
\begin{bmatrix}
u_1' \\
u_2'
\end{bmatrix}
= \begin{bmatrix} 0 \\ 2e^t \end{bmatrix}
\]

\[
\begin{align*}
\begin{bmatrix}
u_1' \\
u_2'
\end{bmatrix}
&= \left( \begin{bmatrix}
e^{-2t} & e^{-3t} \\
-2e^{-2t} & -3e^{-3t}
\end{bmatrix} \right)^{-1}
\begin{bmatrix} 0 \\ 2e^t \end{bmatrix} \\
&= \frac{1}{-e^{-5t}}
\begin{bmatrix}
-3e^{-3t} & -e^{-3t} \\
2e^{-2t} & e^{-2t}
\end{bmatrix}
\begin{bmatrix} 0 \\ 2e^t \end{bmatrix} \\
&= -e^{5t}
\begin{bmatrix}
-2e^{-2t} \\
2e^{-t}
\end{bmatrix}
= \begin{bmatrix}
-2e^{3t} \\
2e^{4t}
\end{bmatrix}
\end{align*}
\]

This tells us, ignoring constants of integration,
\[
u_1(t) = \frac{-2}{3}e^{3t}, \quad \text{and} \quad u_2(t) = \frac{1}{2}e^{4t}.
\]

Hence our particular solution is
\[
Y = u_1\phi_1 + u_2\phi_2 = \frac{-2}{3}e^t + \frac{1}{2}e^t = -\frac{1}{6}e^t.
\]

---

IV.3 • 23. Example 53

Find the general solution to
\[ 4y'' + y = 2\sec(t/2), \quad -\pi < t < \pi. \]

Solution ..

Putting this in the appropriate form, we’re dealing with the differential equation
\[ y'' + \frac{1}{4}y = \frac{1}{2}\sec(t/2). \]

Solutions to the homogeneous version will be
\[ C\cos(t/2) + D\sin(t/2), \]

so let \( \phi_1(t) = \cos(t/2) \) and \( \phi_2(t) = \sin(t/2) \). We know that \( g(t) = \sec(t/2)/2 \). Skipping the steps to find the Wronskian,
\[
W[\phi_1, \phi_2] = \det
\begin{bmatrix}
\phi_1 & \phi_2 \\
\phi_1' & \phi_2'
\end{bmatrix}
= \det
\begin{bmatrix}
\cos(t/2) & \sin(t/2) \\
-\frac{1}{2}\sin(t/2) & \frac{1}{2}\cos(t/2)
\end{bmatrix}
= \frac{1}{2}\cos^2(t/2) + \frac{1}{2}\sin^2(t/2) = \frac{1}{2}.
\]

Thus our formula from variation of parameters yields \( Y = u_1\phi_1 + u_2\phi_2 \) where
\[
u_1 = \int \frac{\phi_2g}{W[\phi_1, \phi_2]} \, dt = \int \frac{\cos(t/2)}{\frac{1}{2}\cos(t/2)} \, dt = \int \frac{1}{2} \tan(t/2) \, dt
\]

\[
u_2 = \int \frac{\phi_2g}{W[\phi_1, \phi_2]} \, dt
\]

\[
u_2 = \int \frac{1}{2} \tan(t/2) \, dt
\]

- \( \int \frac{1}{2} \tan(t/2) \, dt = \int 2 \cos(t/2) \, dt
\]
We can calculate the derivatives, and plug them into the left-hand-side of the differential equation:

\[
\phi_1(x) = \frac{\sin x}{\sqrt{x}}, \quad \phi_2(x) = \frac{\cos x}{\sqrt{x}}
\]

Then we calculate:

\[
\phi_1'(x) = \frac{-\sin x}{2\sqrt{x}} + \frac{\cos x}{\sqrt{x}}, \quad \phi_2'(x) = \frac{-\cos x}{2\sqrt{x^3}} - \frac{\sin x}{\sqrt{x}}
\]

\[
\phi_1''(x) = \frac{3\sin x}{4\sqrt{x^3}} - \frac{\cos x}{\sqrt{x}} - \frac{x^3 \sin x}{4\sqrt{x}}, \quad \phi_2''(x) = \frac{3\cos x}{4\sqrt{x^5}} - \frac{\sin x}{\sqrt{x}} + \frac{x^5 \cos x}{4\sqrt{x^3}}
\]

Thus our particular solution will be, for an interval where \(\cos(t/2) \neq 0\),

\[
Y = 2\ln |\cos(t/2)| \cos(t/2) + t \sin(t/2).
\]

This yields the general solution to the non-homogeneous equation

\[
y(t) = C \cos(t/2) + D \sin(t/2) + \ln |\cos(t/2)| \cos(t/2) + t \sin(t/2)
\]

for constants \(C\) and \(D\).

### IV.3.24. Example 54

Consider the differential equation, for \(x > 0\),

\[
x^2 y'' + xy' + \left(x^2 - \frac{1}{4}\right) y = g(x)
\]

for \(g\) an arbitrary continuous function.

1. Show

\[
\phi_1(x) = \frac{1}{\sqrt{x}} \sin x, \quad \phi_2(x) = \frac{1}{\sqrt{x}} \cos x
\]

are solutions to the homogeneous version.

2. Find the general solution to the non-homogeneous equation.

**Solution:**

1. We can calculate the derivatives, and plug them into the left-hand-side of the differential equation:

\[
\phi_1(x) = -\frac{\sin x}{2\sqrt{x}} + \frac{\cos x}{\sqrt{x}}
\]

\[
\phi_2(x) = \frac{\cos x}{2\sqrt{x^3}} - \frac{\sin x}{\sqrt{x}}
\]

\[
\phi_1''(x) = \frac{3\sin x}{4\sqrt{x^3}} - \frac{\cos x}{\sqrt{x}} - \frac{x^3 \sin x}{4\sqrt{x}}
\]

\[
\phi_2''(x) = \frac{3\cos x}{4\sqrt{x^5}} - \frac{\sin x}{\sqrt{x}} + \frac{x^5 \cos x}{4\sqrt{x^3}}
\]

\[
x^2 y'' + xy' + \left(x^2 - \frac{1}{4}\right) y = \left(3 \sin x \frac{1}{\sqrt{x}} - \sqrt{x} \cos x - \sqrt{x} \sin x\right) + \left(\frac{\sin x}{2\sqrt{x}} + \sqrt{x} \cos x\right)
\]

\[
= \left(3 \sin x \frac{1}{\sqrt{x}} + \frac{-\sin x}{2\sqrt{x}}\right) + \left(\frac{\sin x}{4\sqrt{x}}\right) = 0
\]

\[
x^2 y'' + xy' + \left(x^2 - \frac{1}{4}\right) y = \left(3 \cos x \frac{1}{\sqrt{x}} - \sqrt{x} \cos x + \sqrt{x} \sin x\right) + \left(-\cos x \frac{1}{2\sqrt{x}} - \sqrt{x} \sin x\right)
\]

\[
= \left(3 \cos x \frac{1}{\sqrt{x}} + \frac{-\cos x}{2\sqrt{x}}\right) + \left(-\cos x \frac{1}{\sqrt{x}}\right) = 0
\]

Hence \(\phi_1\) and \(\phi_2\) are solutions to the homogeneous differential equation

\[
x^2 y'' + xy' + \left(x^2 - \frac{1}{4}\right) y = 0.
\]

2. To find the general solution to the non-homogeneous differential equation, proceed by variation of parameters. First we calculate the Wronskian of \(\phi_1\) and \(\phi_2\). Since \(x > 0\), this will be non-zero:

\[
W[\phi_1, \phi_2] = \det \begin{bmatrix} \frac{\phi_1'}{\phi_1} & \frac{\phi_2}{\phi_1} \\ \frac{\phi_1'}{\phi_1} & \frac{\phi_2}{\phi_2} \end{bmatrix} = \det \begin{bmatrix} \frac{\sin x}{\sqrt{x}} & \cos x \frac{1}{\sqrt{x}} \cos x \frac{1}{\sqrt{x}} \end{bmatrix}
\]
Hence \( \phi_1 \) and \( \phi_2 \) form a fundamental set of solutions, and so we get a particular solution of the form

\[
Y = u_1 \phi_1 + u_2 \phi_2
\]

where

\[
u_1 = -\int \frac{\phi_2 g}{W[\phi_1, \phi_2]} \, dx \quad \text{and} \quad u_2 = \int \frac{\phi_2 g}{W[\phi_1, \phi_2]} \, dx
\]

So the general solution is

\[
y = C \sin x + D \cos x + Y
\]

for \( Y \) defined as above, and constants \( C \) and \( D \).

---

**IV.3 • 25. Example 55**

Solve the equation \( y'' + y' + 10y = 0 \); with \( y(0) = 1, \ y'(0) = 0 \).

**Solution:**

\( e^{rt} \) is a solution to this equation if

\[
e^{rt}(r^2 + r + 10) = 0 \quad \iff \quad r^2 + r + 10 = 0
\]

which yields

\[
r = \frac{-1 \pm i\sqrt{39}}{2}.
\]

Call these two values \( r_1 \) and \( r_2 \). The Wronskian of \( e^{r_1 t} \) and \( e^{r_2 t} \) will not be 0 since \( r_1 \neq r_2 \). Thus any solution will be of the form

\[
y(t) = Ce^{r_1 t} + De^{r_2 t}
\]

for constants \( C \) and \( D \). As complex numbers, we get that this is equivalent to

\[
y(t) = C_1 e^{-t/2} \cos(t\sqrt{39}/2) + C_2 e^{-t/2} \sin(t\sqrt{39}/2)
\]

for constants \( C_1 \) and \( C_2 \).

The initial condition \( y(0) = 1 \) implies \( 1 = C_1 + 0 \). Differentiating, and writing \( A \) for \( \sqrt{39}/2 \),

\[
y'(t) = -C_1 \frac{1}{2} e^{-t/2} \cos(At) - CAe^{-t/2} \sin(At) - C_2 \frac{1}{2} e^{-t/2} \sin(At) + C_2 A e^{-t/2} \cos(At)
\]

\[
= -C_1 \frac{1}{2} + 0 + 0 + C_2 A
\]

\[
\therefore C_2 = \frac{1}{\sqrt{39}}.
\]

Thus we know both \( C_1 \) and \( C_2 \), which tells us \( y(t) \):

\[
y(t) = e^{-t/2} \cos(t\sqrt{39}/2) + \frac{1}{\sqrt{39}} e^{-t/2} \sin(t\sqrt{39}/2).
\]
Solve the initial value problem

\[ y'' + y' + 5y = t - e^{-t}, \]

where \( y(0) = 0 \), and \( y'(0) = 0 \).

**Solution:**

As with any linear, non-homogeneous, differential equation, proceed by first solving the homogeneous version. In this case, this means solving \( y'' + y' + 5y = 0 \). This has solutions of the form \( e^{rt} \) whenever \( e^{rt}(r^2 + r + 5) = 0 \), i.e. \( r = -1/2 \pm i\sqrt{19}/2 \). Using previous results, Euler’s formula, and taking new arbitrary constants, the general solution to the homogeneous version is then

\[ C\phi_1 + D\phi_2 = Ce^{-t/2} \sin(\sqrt{19}t/2) + De^{-t/2} \cos(\sqrt{19}t/2) \]

for constants \( C \) and \( D \). Now we must solve the non-homogeneous equation for a particular solution \( Y(t) \). To do this, proceed by undetermined coefficients. The function \( t - e^{-t} \) is the sum of a first degree polynomial, and an exponential, so our guess will be the sum of the corresponding guesses:

\[
\begin{align*}
Y(t) &= At + B + Ge^{-t} \\
Y'(t) &= A - Ge^{-t} \\
Y''(t) &= Ge^{-t}.
\end{align*}
\]

Thus for \( Y(t) \) to solve the differential equation, we require

\[
\begin{align*}
t - e^{-t} &= Y'' + Y' + 5Y \\
&= Ge^{-t} + A - Ge^{-t} + 5At + 5B + 5Ge^{-t} \\
&= 5At + 5Ge^{-t} + (A + 5B) \\
&\iff 5A = 1, \text{ and } 5G = -1, \text{ and } A + 5B = 0 \\
&\iff A = \frac{1}{5}, \text{ and } G = -\frac{1}{5}, \text{ and } B = -\frac{1}{25}.
\end{align*}
\]

By equation coefficients, this requires \( 5A = 1, 5G = -1, \) and \( A + 5B = 0 \). In other words, \( A = 1/5, B = -1/25, \) and \( G = -1/5 \). Hence we get a particular solution \( Y(t) = t/5 - 1/25 - e^{-t}/5 \). With the homogeneous part, the general solution to the non-homogeneous, differential equation is, writing \( \alpha \) for \( \sqrt{19}/2 \),

\[
\begin{align*}
y(t) &= Ce^{t/2} \sin(\alpha t) + De^{-t/2} \cos(\alpha t) + \frac{t}{5} - \frac{1}{25} - \frac{1}{5}e^{-t} \\
y'(t) &= \frac{2C\alpha - D}{2}e^{-t/2} \cos(\alpha t) - \frac{C + 2D\alpha}{2}e^{-t/2} \sin(\alpha t) + \frac{1}{5} + \frac{1}{5}e^{-t}.
\end{align*}
\]

So using the initial conditions requires

\[
\begin{align*}
0 &= D - \frac{1}{25} - \frac{1}{5} \\
0 &= C\alpha - \frac{D}{2} + \frac{2}{5},
\end{align*}
\]

\[ \therefore D = \frac{6}{25}, \text{ and } C = -\frac{14}{25\sqrt{19}}. \]

Hence the solution is

\[
y(t) = \frac{14}{25\sqrt{19}} e^{-t/2} \sin \left( \frac{\sqrt{19}}{2} t \right) + \frac{6}{25} e^{-t/2} \cos \left( \frac{\sqrt{19}}{2} t \right) + \frac{t}{5} - \frac{1}{25} - \frac{1}{5}e^{-t}.
\]
IV.3.27. Example 57

Solve the equation

\[ y'' + 3y = \begin{cases} 
  t & \text{if } 0 \leq t \leq 1 \\
  1 & \text{if } 1 < t,
\end{cases} \]

where \( y(0) = 1 \), and \( y'(0) = 1 \).

**Proof.**

For the sake of space, write

\[ g(t) := \begin{cases} 
  t & \text{if } 0 \leq t \leq 1 \\
  1 & \text{if } 1 < t.
\end{cases} \]

As with any linear, non-homogeneous, differential equation, proceed by first solving the homogeneous version. In this case, this means solving \( y'' + 3y = 0 \). This has solutions of the form \( e^{rt} \) whenever \( e^{rt}(r^2 + 3) = 0 \), i.e. whenever \( r = \pm i\sqrt{3} \). Using previous results, Euler’s formula, and taking new constants, the general solution to the homogeneous version is

\[ C\phi_1 + D\phi_2 = C\sin(\sqrt{3}t) + D\cos(\sqrt{3}t), \]

for some constants \( C \) and \( D \). Now we must solve the non-homogeneous equation for some particular solution. We do this by variation of parameters. At this point, we have the following things set:

\[ \phi_1 = \sin(\sqrt{3}t) \]
\[ \phi_2 = \cos(\sqrt{3}t) \]
\[ g(t) = \begin{cases} 
  t & \text{if } 0 \leq t \leq 1 \\
  1 & \text{if } 1 < t,
\end{cases} \]
\[ W[\phi_1, \phi_2] = \det \begin{bmatrix} \phi_1 & \phi_2 \\
 \phi_1' & \phi_2'
\end{bmatrix} = \det \begin{bmatrix} \sin(\sqrt{3}t) & \cos(\sqrt{3}t) \\
 \sqrt{3}\cos(\sqrt{3}t) & -\sqrt{3}\sin(\sqrt{3}t)
\end{bmatrix}
\]
\[ = -\sqrt{3}\sin^2(\sqrt{3}t) - \sqrt{3}\cos^2(\sqrt{3}t)
\]
\[ = -\sqrt{3} \neq 0. \]

Variation of parameters then says that \( Y(t) := u_1(t)\phi_1(t) + u_2(t)\phi_2(t) \) is a particular solution to the non-homogeneous equation for \( u_1 \) and \( u_2 \) defined below.

\[ u_1(t) = \int_0^t \frac{-\phi_2(x)g(x)}{W[\phi_1, \phi_2]} \, dx = \frac{1}{\sqrt{3}} \int_0^t \cos(\sqrt{3}x)g(x) \, dx 
\]
\[ u_2(t) = \int_0^t \frac{\phi_1(x)g(x)}{W[\phi_1, \phi_2]} \, dx = -\frac{1}{\sqrt{3}} \int_0^t \sin(\sqrt{3}x)g(x) \, dx. \]

Given that \( t \) might be between 0 and 1, or \( t > 1 \), we have two cases to consider when we look at the bounds of integration above. What this means is that we will parse this integral into two parts depending on \( t \):\n
\[ u_1(t) = \frac{1}{\sqrt{3}} \int_0^{t} \cos(\sqrt{3}x) \, dx \quad \text{if } 0 \leq t \leq 1, \]
\[ u_1(t) = \frac{\cos(\sqrt{3})}{\sqrt{3}} \int_0^{1} \cos(\sqrt{3}x) \, dx + \frac{1}{\sqrt{3}} \int_1^{t} \cos(\sqrt{3}x) \, dx \quad \text{if } t > 1. \]

We get similar equalities for \( u_2 \). Evaluating these integrals on their respective bounds, we find after simplifying that \( Y(t) \) is given by

\[ Y(t) = \begin{cases} 
  \frac{t}{3} - \frac{1}{3\sqrt{3}} \sin(\sqrt{3}t) & \text{if } 0 \leq t \leq 1 \\
  \frac{1}{3} + \frac{\cos(\sqrt{3}) - 1}{3\sqrt{3}} \sin(\sqrt{3}t) - \frac{\sin(\sqrt{3})}{3\sqrt{3}} \cos(\sqrt{3}t) & \text{if } 1 \leq t.
\end{cases} \]

Note that \( Y(t) \) is twice differentiable on from 0 to \( \infty \), including \( t = 1 \). So our general solution to the differential
Chapter IV – Section 3

The solution is

\[ y = C\phi_1 + D\phi_2 + Y \]

for constants \( C \) and \( D \). Using the initial condition of \( y(0) = 1 \), we get that \( 1 = C \cdot 0 + D \cdot 1 + 0 = D \). Differentiating and using the initial condition of \( y'(0) = 1 \) yields that \( y'(0) = 1 = \sqrt{3}C \), i.e. \( C = 1/\sqrt{3} \). Hence we can write that the solution is

\[ y(t) = \begin{cases} \frac{2}{3\sqrt{3}} \sin(\sqrt{3}t) + \cos(\sqrt{3}t) + \frac{1}{3} & \text{if } 0 \leq t \leq 1 \\ \frac{\cos(\sqrt{3}t) + 2}{3\sqrt{3}} \sin(\sqrt{3}t) + \left(1 - \frac{\sin(\sqrt{3})}{3\sqrt{3}}\right)\cos(\sqrt{3}t) + \frac{1}{3} & \text{if } 1 < t. \end{cases} \]

There is another way to do this problem, which is by undetermined coefficients. To do this, however, we would need to do it in an unusual way. The homogeneous solution, rather than \( C\phi_1 + D\phi_2 \) like usual, needs to have separate constants over each interval. In essence, we must solve the equation

\[ y'' + 3y = t \]

and then the equation

\[ y'' + 3y = 1, \]

and then glue the two solutions together at \( t = 1 \). The next solution to the same problem gives this method.

### IV.3.28. Example 57

Solve the equation

\[ y'' + 3y = \begin{cases} t & \text{if } 0 \leq t < 1 \\ 1 & \text{if } 1 < t. \end{cases} \]

where \( y(0) = 1 \), and \( y'(0) = 1 \).

**Solution:**

For the sake of space, write

\[ g(t) := \begin{cases} t & \text{if } 0 \leq t < 1 \\ 1 & \text{if } 1 < t. \end{cases} \]

As with any linear, non-homogeneous, differential equation, proceed by first solving the homogeneous version. In this case, this means solving \( y'' + 3y = 0 \). This has solutions of the form \( e^{rt} \) whenever \( e^{rt}(r^2 + 3) = 0 \), i.e. whenever \( r = \pm i\sqrt{3} \). Using previous results, Euler’s formula, and taking new constants, the general solution to the homogeneous version is

\[ C\phi_1 + D\phi_2 = C\sin(\sqrt{3}t) + D\cos(\sqrt{3}t), \]

for some constants \( C \) and \( D \). Now we can solve two separate differential equations.

1. First we solve \( y'' + 3y = t \) for \( y \) with initial value \( y(0) = 1 \) and \( y'(0) = 1 \). The solution will be

\[ y_1(t) = C_1 \sin(\sqrt{3}t) + D_1 \cos(\sqrt{3}t) + Y_1(t) \]

for some \( Y_1(t) \). Proceeding by undetermined coefficients, we will guess \( Y_1(t) = At + B \), which yields that \( A = 1/3 \), and \( B = 0 \). Hence the first solution is

\[ y_1(t) = C_1 \sin(\sqrt{3}t) + D_1 \cos(\sqrt{3}t) + \frac{1}{3}t. \]

The initial value of \( y_1(0) = 1 \) gives that \( D_1 = 1 \). By differentiation, the initial condition \( y'_1(0) = 1 \) gives that \( \sqrt{3}C_1 + \frac{1}{3} = 1 \), i.e. \( C_1 = 2/(3\sqrt{3}) \). Thus our first solution is

\[ y_1(t) = \frac{2}{3\sqrt{3}} \sin(\sqrt{3}t) + \cos(\sqrt{3}t) + \frac{1}{3}t. \]

2. The next solution solves \( y'' + 3y = 1 \) for \( y_2 \) with initial conditions

\[ y_2(1) = y_1(1) = \frac{2}{3\sqrt{3}} \sin(\sqrt{3}) + \cos(\sqrt{3}) + \frac{1}{3} \]
To do this, we again proceed by undetermined coefficients with a guess of $y_2(t) = At + B$. From this, $Y''(t) = 0$, and so we get $A = 0$, and $B = 1/3$. Hence, with the same general solution to the homogeneous version, we get

$$y_2(t) = C_2 \sin(\sqrt{3}t) + D_2 \cos(\sqrt{3}t) + \frac{1}{3}.$$  

The first initial condition allows us to say

$$y_2(1) = y_1(1) = \frac{2}{3} \cos(\sqrt{3}) - \sqrt{3} \sin(\sqrt{3}) + \frac{1}{3} = C_2 \sin(\sqrt{3}) + D_2 \cos(\sqrt{3}) + \frac{1}{3}.$$  

The second initial condition tells us

$$y_2'(1) = \frac{2}{3} \cos(\sqrt{3}) - \sqrt{3} \sin(\sqrt{3}) + \frac{1}{3} = y_2'(1) = \sqrt{3}C_2 \cos(\sqrt{3}) - \sqrt{3}D_2 \sin(\sqrt{3}).$$  

Solving these and doing a lot of simplification yields that

$$C_2 = \frac{\cos(\sqrt{3}) + 2}{3\sqrt{3}}, \quad \text{and} \quad D_2 = 1 - \frac{\sin(\sqrt{3})}{3\sqrt{3}}.$$  

which gives the same answer as with variation of parameters:

$$y(t) = \begin{cases} C_1 \phi_1 + D_1 \phi_2 + Y_1 & \text{if } 0 \leq t \leq 1 \\ C_2 \phi_2 + D_2 \phi_2 + Y_2 & \text{if } 1 < t \end{cases}$$

$$= \begin{cases} \frac{2}{3\sqrt{3}} \sin(\sqrt{3}t) + \cos(\sqrt{3}t) + \frac{t}{3} & \text{if } 0 \leq t \leq 1 \\ \frac{\cos(\sqrt{3})+2}{3\sqrt{3}} \sin(\sqrt{3}t) + \left(1 - \frac{\sin(\sqrt{3})}{3\sqrt{3}}\right) \cos(\sqrt{3}t) + \frac{1}{3} & \text{if } 1 < t. \end{cases}$$

---

**IV.3 • 29. Example 58**

Solve the equation

$$x^2 \frac{d^2 y}{dx^2} + 2x \frac{dy}{dx} + 3y = 0$$

with initial conditions $y(2) = 1$ and $y'(2) = 0$ for $x \geq 2$.

**Solution .:.**

First manipulate the quation to be of the form

$$\frac{d^2 y}{dx^2} + \frac{2}{x} \frac{dy}{dx} + \frac{3}{x^2} y = 0.$$  

Consider the variable $t = \ln x$, which is well-defined since $x \geq 2$. This gives $x'(t) = x(t) = e^t$. We then have by the chain rule that

$$\frac{dy}{dx} = \frac{dy}{dt} \frac{dt}{dx} = \frac{dy}{dt} \frac{1}{x},$$

$$\frac{d^2 y}{dx^2} = \frac{d}{dt} \left( \frac{dy}{dt} \frac{1}{x} \right) = \frac{d^2 y}{dt^2} \frac{1}{x} - \frac{dy}{dt} \frac{1}{x^2}.$$  

substituting this into the above equation yields the new differential equation

$$\frac{d^2 y}{dt^2} - \frac{dy}{dt} + 2 \frac{dy}{dt} + 3y = 0 \iff \frac{d^2 y}{dt^2} + \frac{dy}{dt} + 3y = 0.$$  

---

63
solving this can be done like any other second-order, homogeneous, differential equation. We get solutions of the form $e^{rt}$ for $r$ satisfying

$$r^2 + r + 3 = 0,$$

meaning $r = (-1 \pm i \sqrt{11})/2$. Hence the general solution (in terms of $t$) is, writing $\alpha$ for $\sqrt{11}/2$,

$$y(t) = Ce^{-t/2} \sin(\alpha t) + De^{-t/2} \cos(\alpha t),$$

$$y'(t) = \frac{-y(t)}{2} - \alpha Ce^{-t/2} \cos(\alpha t) - \alpha De^{-t/2} \sin(\alpha t).$$

for constants $C$ and $D$. The initial conditions in terms of $t$ are that $y(ln 2) = 1$, and $y'(ln 2) = 0$. This translates to the linear equations, writing $\alpha$ for $\sqrt{11}/2$, and noting that $e^{-ln(2)/2} = 1/\sqrt{2}$,

$$1 = \frac{C}{\sqrt{2}} \sin(\alpha \ln(2)) + \frac{D}{\sqrt{2}} \cos(\alpha \ln(2))$$

$$0 = -\frac{1}{2} + \frac{\alpha C}{\sqrt{2}} \cos(\alpha \ln(2)) - \frac{\alpha D}{\sqrt{2}} \sin(\alpha \ln(2))$$

$$\longleftrightarrow \frac{1}{2} = \frac{\alpha C}{\sqrt{2}} \cos(\alpha \ln(2)) - \frac{\alpha D}{\sqrt{2}} \sin(\alpha \ln(2))$$

Solving for $C$ and $D$ yields the expressions

$$C = \sqrt{2} \sin(\alpha \ln(2)) + \frac{1}{\sqrt{2\alpha}} \cos(\alpha \ln(2))$$

$$D = \sqrt{2} \cos(\alpha \ln(2)) - \frac{1}{\sqrt{2\alpha}} \sin(\alpha \ln(2)).$$

substituting this back into the equation for $y$ in terms of $t$ and substituting $t$ for $e^x$ gives the final answer for $y$ in terms of $x$: after simplification, still writing $\alpha$ for $\sqrt{11}/2$,

$$y(x) = \sqrt{\frac{2}{\alpha}} \left( \cos(\alpha \ln(2)) + \frac{\sin(\alpha \ln(2))}{2\alpha} \right) \cos(\alpha \ln(x))$$

$$+ \sqrt{\frac{2}{\alpha}} \left( \sin(\alpha \ln(2)) + \frac{\cos(\alpha \ln(2))}{2\alpha} \right) \sin(\alpha \ln(x)).$$
Chapter V. Using Linear Algebra

Section V.1. Gaussian Elimination

§ V.1.A. Motivating Gaussian elimination

Our goal here is just to go over Gaussian elimination, a process to invert a matrix. Our setup will be equations of the form $A \mathbf{v} = \mathbf{w}$ for a given matrix $A$, and vectors $\mathbf{v}$ and $\mathbf{w}$. In principle, we would like to say $\mathbf{v} = A^{-1} \mathbf{w}$, but not all matrices can be inverted, and even when they can, finding the inverse is sometimes a difficult task. Gaussian elimination is a procedure to find $\mathbf{v}$ given $\mathbf{w}$ and $A$, and in essence, find such an inverse if it exists.

This process can be useful for us in differential equations when looking at higher-order differential equations—i.e. third-order, fourth-order, etc. In particular, linear, non-homogeneous, differential equations require us to solve the matrix equation

$$W \cdot \begin{bmatrix} u'_1 \\ \vdots \\ u'_{n-1} \\ u'_n \end{bmatrix} \begin{bmatrix} \phi'_1 \\ \vdots \\ \phi'_{n-1} \\ \phi'_n \end{bmatrix} = \begin{bmatrix} \phi_1 \\ \vdots \\ \phi_{n-1} \\ \phi_n \end{bmatrix} \begin{bmatrix} u'_1 \\ \vdots \\ u'_{n-1} \\ u'_n \end{bmatrix} = \begin{bmatrix} 0' \\ \vdots \\ 0' \end{bmatrix},$$

where $W$ is the Wronskian matrix. Solving this gives $Y = u_1 \phi_1 + \cdots + u_n \phi_n$ as a particular solution: the $n$th-order version of variation of parameters. Things get more complicated especially when we consider systems of differential equations. For more information about this, consider Chapter VI.

§ V.1.B. Motivation from systems of equations

Recall that with just a couple linear equations

$Ax + By = N$

$Cx + Dy = M,$

we can solve the equation by adding and subtracting the respective equalities from each other, yielding the following as an example.

<table>
<thead>
<tr>
<th>V.1.B • 1. Example 59</th>
</tr>
</thead>
<tbody>
<tr>
<td>Consider the system of equations</td>
</tr>
<tr>
<td>$-x + 3y = 5$</td>
</tr>
<tr>
<td>$2x + 5y = 10.$</td>
</tr>
</tbody>
</table>

Procedure .:

Adding “twice the first equation” to the second equation means that we can conclude

$$(2x + 5y) + 2(-x + 3y) = 10 + 2 \cdot 5$$

$$11y = 20 \rightarrow y = 20/11.$$  

Using this result, the first equation then says $-x + 60/11 = 5$, i.e. $x = 5/11.$
This can allow us to quickly solve the equations if the coefficients are nice enough like here. Gaussian elimination is just reformulating this idea in the context of matrices.

**Matrix Procedure:**

We could instead write the setup for this example as the matrix equation

\[
\begin{bmatrix}
1 & 3 \\
2 & 5
\end{bmatrix}
\begin{bmatrix}
x \\
y
\end{bmatrix}
=
\begin{bmatrix}
5 \\
10
\end{bmatrix}.
\]

the idea of “adding twice the first equation to the second” just means multiplying by the matrix

\[
\begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
0 & 0 \\
2 & 0
\end{bmatrix}
=\begin{bmatrix}
1 & 0 \\
2 & 1
\end{bmatrix}.
\]

This transforms our matrix equation into

\[
\begin{bmatrix}
1 & 0 \\
2 & 1
\end{bmatrix}\begin{bmatrix}
-1 & 3 \\
2 & 5
\end{bmatrix}\begin{bmatrix}
x \\
y
\end{bmatrix}
=\begin{bmatrix}
1 & 0 \\
2 & 1
\end{bmatrix}\begin{bmatrix}
5 \\
10
\end{bmatrix}.
\]

which expresses the same equality as before: $11y = 20$. We can then multiply by a scaling matrix, and so yield the equality below.

\[
\begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
1 & 3 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
x \\
y
\end{bmatrix}
=\begin{bmatrix}
0 & 0 \\
2 & 1
\end{bmatrix}\begin{bmatrix}
1 & 0 \\
2 & 1
\end{bmatrix}\begin{bmatrix}
5 \\
10
\end{bmatrix}.
\]

adding thrice the second row to the first then yields

\[
\begin{bmatrix}
1 & 3 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
1 & -3 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
x \\
y
\end{bmatrix}
=\begin{bmatrix}
0 & 0 \\
1 & 1
\end{bmatrix}\begin{bmatrix}
1 & 3 \\
0 & 1
\end{bmatrix}\begin{bmatrix}
-5 \\
20/11
\end{bmatrix}.
\]

In other words, $x = 5/11$, and $y = 20/11$.

This has just been motivation, however. We still need to explicitly lay out the process.

### § V.1.C. Setting up Gaussian elimination

Our goal is to manipulate the matrix equation

\[
\begin{bmatrix}
a_{11} & \cdots & a_{1n} \\
\vdots & \ddots & \vdots \\
a_{n1} & \cdots & a_{nn}
\end{bmatrix}\begin{bmatrix}
x_1 \\
\vdots \\
x_n
\end{bmatrix}
=\begin{bmatrix}
b_1 \\
\vdots \\
b_n
\end{bmatrix}
\]

for some $c_1, \cdots, c_n$. Note that this equation on the right is just saying $\mathbf{x} = \mathbf{c}$. We do this by applying three operations multiple times. All of these operations are just multiplying both sides of the equation by an invertible matrix on the left.

### V.1.C • 1. Definition 16

An elementary row operation on a matrix is an operation that

1. scales a row;
2. adds (a multiple of) one row to another; or
3. switches two rows.

Note that these operations can be accomplished by multiplying by a certain invertible matrix on the left. This allows us to transform matrices into other matrices which are much nicer to work with. In particular, we will care about reduced row echelon form.
An entry \( a \) in a matrix \( A \) is a leading coefficient iff it is the first non-0 entry in its row of \( A \).

A matrix \( A \) is in reduced row echelon form iff

1. every leading coefficient of \( A \) is 1;
2. every leading coefficient has only 0s in its column;
3. every row’s leading coefficient appears to the right of higher rows’; and
4. rows with only 0s appear at the bottom.

All the following are examples of matrices in reduced row echelon form (‘\( \ast \)’ just represents some entry that doesn’t need to be 0)

\[
\begin{bmatrix}
1 & \ast & 0 & \ast & \ast & 0 \\
0 & 0 & 1 & \ast & \ast & 0 \\
0 & 0 & 0 & 0 & 0 & 1
\end{bmatrix},
\begin{bmatrix}
1 & 0 & \ast & \ast \\
0 & 1 & \ast & \ast \\
0 & 0 & 0 & 0 & 1
\end{bmatrix},
\begin{bmatrix}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{bmatrix}.
\]

All the following are examples of matrices that are not in reduced row echelon form.

\[
\begin{bmatrix}
1 & \ast & \ast & \ast & \ast & 0 \\
0 & 1 & \ast & \ast & \ast & 0 \\
0 & 0 & 0 & 0 & 0 & 1
\end{bmatrix},
\begin{bmatrix}
0 & 1 & \ast & \ast \\
1 & 0 & \ast & \ast \\
0 & 0 & 0 & 1
\end{bmatrix},
\begin{bmatrix}
1 & \ast & 0 \\
0 & 0 & 0 \\
0 & 0 & 1
\end{bmatrix}.
\]

The first isn’t since row 2’s leading coefficient has a non-0 entry right above it. The second isn’t because the second row’s leading coefficient appears to the left of the first row’s. The third isn’t because the there is a row of 0s which doesn’t appear at the bottom.

This is all the background we need to get actually define the method of Gaussian elimination.

### § V.1.D. The method of Gaussian elimination

So this is the setup to Gaussian elimination: we have a matrix equation \( Ax = b \), and we want to transform \( A \) into reduced row echelon form by way of elementary row operations.

Consider the equation \( A \cdot x = b \) for a given matrix \( A \), given vector \( b \), and vector of variables \( x \).

Gaussian elimination is the process of transforming the matrix \([A|b]\)—just the entries of \( A \) next to the entries of \( b \)—by applying elementary row operations into reduced row echelon form.

Often it’s useful to instead do the process on a different matrix that combines \( A \) and \( b \). Looking back to Example 59 (V.1.B • 1), we can again go through the process, starting with

\[
\begin{bmatrix}
-1 & 3 & 5 \\
2 & 5 & 10
\end{bmatrix}
\]

and doing the same operations as before:

\[
\begin{bmatrix}
-1 & 3 & 5 \\
2 & 5 & 10
\end{bmatrix} \rightarrow \begin{bmatrix}
-1 & 3 & 5 \\
0 & 1 & 11/20
\end{bmatrix} \rightarrow \begin{bmatrix}
1 & -3 & -5 \\
0 & 1 & 20/11
\end{bmatrix} \rightarrow \begin{bmatrix}
1 & 0 & 5/11 \\
0 & 1 & 20/11
\end{bmatrix}.
\]

Very useful is the following result, which tells us that—if we can invert the matrix—Gaussian elimination takes us all the way to the answer we want: it solves the equation.

\(^1\)going from left to right
V.1.D • 2. Result 22 (Gaussian elimination on invertible matrices)

Let $A$ be an invertible matrix, and $b$ a vector. Consider the equation $A \cdot x = b$, where $x$ is a vector of variables.

Let $R$ be the result of doing Gaussian elimination to $A$, and $\vec{c}$ the result of this process applied to $b$.

Thus $R$ is the identity matrix, and $x = \vec{c}$.

\textit{Proof} ::

It turns out (unproven here) that the reduced row echelon form of a matrix is unique in the sense that there is only one matrix $P$ such that some non-zero matrix $M$ has $M \cdot A = P$ in reduced row echelon form.

Given that $A$ is invertible, $A^{-1} \cdot A = \text{Id}$ is in reduced row echelon form, and so $R = \text{Id}$. So the process of doing Gaussian elimination is the same as applying $A^{-1}$. Thus $\vec{c} = A^{-1}b = x$. ⊤

If we can invert the matrix, we have a pretty simple idea to solve the equation. The idea generalizes easily from the $3 \times 3$ case. Suppose we have the augmented matrix

\[
\begin{bmatrix}
* & * & * \\
* & * & * \\
* & * & * \\
\end{bmatrix}
\]

First, we just scale the first row so that the first entry becomes 1:

\[
\begin{bmatrix}
1 & * & * \\
* & * & * \\
* & * & * \\
\end{bmatrix}
\]

Now we subtract the appropriate amount of row 1 from the other two rows to clear out the first column:

\[
\begin{bmatrix}
1 & * & * \\
0 & * & * \\
0 & * & * \\
\end{bmatrix}
\]

Now we do the same thing for the second column, working with the second row:

\[
\begin{bmatrix}
1 & * & * \\
0 & * & * \\
\end{bmatrix} \mapsto \begin{bmatrix}
1 & * & * \\
0 & 1 & * \\
\end{bmatrix} \mapsto \begin{bmatrix}
1 & 0 & * \\
0 & 1 & * \\
\end{bmatrix}
\]

and then the same for the third. The general process is to scale to 1, and then clear out the rest of the column. This allows us to work with the other columns more easily. In a similar way, switching rows can be helpful, like if there is already a 1 in the first column. Switching that row and the first row would put a 1 in the top left, and so you wouldn’t need to scale. Similarly, if just through a simple addition/subtraction of a row, we can get a 1, that may be easier than scaling. Ultimately, we’ll arrive at the same answer, but the computations involved are already tedious, and involving fractions can make it more-so, so there’s a bit of a disincentive to scale if we don’t have to. We also don’t need to start with the first column: we can do them in any order.

Note that we can also generalize this process to find the inverse of a matrix directly (if there is an inverse). In particular, we can find the first column of the inverse matrix by solving

\[A \cdot x = [1 \ 0 \ \cdots \ 0]^\top.\]

Similarly, we can find the second column of the inverse matrix, and then the third, fourth, etc. through the same sort of process, solving for $x$ in each of these. But note that we can absorb this iterative process into just one process, solving

\[A \cdot X = \text{Id}.\]

So instead of starting with many augmented matrices $[A \mid e]$ for some vector $e$—a bunch of 0s and one 1—we can start with the augmented matrix below, and go as before to turn this into reduced row echelon form; turning $A$ into $\text{Id}$.

\[
[A \mid \text{Id}] = \begin{bmatrix}
a_{11} & \cdots & a_{1n} & 1 & 0 \\
\vdots & \ddots & \vdots & \ddots & \vdots \\
a_{n1} & \cdots & a_{nn} & 0 & 1
\end{bmatrix}.
\]
Section V.2. Matrices

There are a few fundamental operations on matrices we will be looking at, namely multiplication, addition, scaling, transposition, and conjugation. These are defined as follows.

Consider the matrix $A$, where the $i$, $j$th slot is $a_{ij}$. This is written $A = (a_{ij})_{i,j}$. Here $i$ refers to the row, and $j$ to the column.

Consider the matrices $A = (a_{ij})_{i,j}$, and $B = (b_{ij})_{i,j}$. Let $k$ be a constant.

1. $A + B = (a_{ij} + b_{ij})_{i,j}$, i.e. the matrix where we add the corresponding entries.
2. $kA = (ka_{ij})_{i,j}$, i.e. the matrix where we scale all the entries.
3. $A \cdot B$ is given by the following:
   
   $$A \cdot B = \left( \sum_k a_k b_{kj} \right)_{i,j}.$$

4. $A^\top$, the transpose, is $(a_{ji})_{i,j}$.

5. $\overline{A}$, the conjugate, is $(\overline{a_{ij}})_{i,j}$, taking the complex conjugate of all the entries.
6. $A^* = (A^\top)^\top = (\overline{A})^\top$, taking the conjugate and transposing (in either order).

Matrix multiplication can be understood as successive multiplication of columns and rows. Consider the equality

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & j \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} ax + by + cz \\ dx + ey + fz \\ gx + hy + jz \end{bmatrix}.$$

We can then define matrix multiplication of a full matrix by doing this row by row on the left:

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & j \end{bmatrix} \cdot \begin{bmatrix} x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{bmatrix} = \begin{bmatrix} ax_1 + by_1 + cz_1 & ax_2 + by_2 + cz_2 & ax_3 + by_3 + cz_3 \\ dx_1 + ey_1 + fz_1 & dx_2 + ey_2 + fz_2 & dx_3 + ey_3 + fz_3 \end{bmatrix}.$$

Now we can define full matrix multiplication by doing this column by column on the right:

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & j \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ y_1 \\ z_1 \\ x_2 \\ y_2 \\ z_2 \\ x_3 \\ y_3 \\ z_3 \end{bmatrix} = \begin{bmatrix} ax_1 + by_1 + cz_1 & ax_2 + by_2 + cz_2 & ax_3 + by_3 + cz_3 \\ dx_1 + ey_1 + fz_1 & dx_2 + ey_2 + fz_2 & dx_3 + ey_3 + fz_3 \end{bmatrix}.$$

Matrix multiplication can be understood as composition, though this isn’t so important for us. For example, suppose $a = 2x - y$ and $b = 5x + y$, and $v = a + b$, $w = -2a - b$. These are represented by the matrix equations

$$\begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 5 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix}, \quad \begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \end{bmatrix}.$$

If we wanted to write $v$ and $w$ in terms of $x$ and $y$ instead of $a$ and $b$, we would get

$$v = (2x - y) + (5x + y) = 7x + 0y, \quad w = -2(2x - y) - (5x + y) = -9x + y,$$

which, when represented as a matrix equation, is precisely what we get when we multiply the matrices:

$$\begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \cdot \begin{bmatrix} 2 & -1 \\ 5 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ 7 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ y \end{bmatrix}.$$

Inverting a matrix is also a kind of operation, but only exists for certain matrices, namely those which are invertible. This has been covered in Section V.1, where Gaussian elimination is a method to find the inverse of a matrix.

We have one last definition for looking at matrices: the determinant. The determinant is very useful mostly for its
properties, which actually determine the function. Note also that the determinant is only defined for square matrices. Below are just some properties of the determinant.

V.2.2. Result 23

For square matrices $A$ and $B$, of size $n \times n$,

\[
\begin{align*}
\det(\text{Id}) &= 1 \\
\det(A) &= a_{11} \cdot a_{22} \cdot a_{33} \cdot \ldots \cdot a_{nn}, \text{ if } A \text{ is triangular}, \\
\det(A^\top) &= \det(A) \\
\det(A \cdot B) &= \det(A) \cdot \det(B) \\
\det(cA) &= c^nA \\
\det(A) &= 0 \text{ if two rows or columns of } A \text{ are the same.}
\end{align*}
\]

The precise definition of the determinant is given below.

V.2.3. Definition 20

Let $A = (a_{ij})_{i,j}$ be an $n \times n$ matrix. The determinant of $A$ is given by

\[
\det(A) = \sum_{i_1, j_1=1}^{n} \pm a_{i_1,j_1} \cdots a_{i_n,j_n},
\]

where the sign of $\pm$ is given by the indices. In other words, we sum up all products of elements, where only one element from each row and column is chosen at a time, and either add or subtract this.

This hasn’t actually defined the determinant completely, so the following are the determinants of $2 \times 2$ and $3 \times 3$ matrices in general.

\[
\begin{align*}
\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} &= ad - bc \\
\det \begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{bmatrix} &= a_1(b_2c_3 - c_2b_3) + b_1(c_2a_3 - a_2c_3) + c_1(a_2b_3 - b_2a_3). 
\end{align*}
\]

Section V.3. Linear Algebra

§ V.3.A. Linear independence

First we have the notion of linear independence.

\footnote{Just like how integration on continuous functions is uniquely defined by agreeing with the area of rectangles, and adding over disjoint intervals.}

\footnote{Almost everything in this section is discussing finite dimensional vector spaces. Generalizing to infinite dimensional ones requires slight changes to all this terminology in the way that I’ve written it here.}
V.3.A • 1. Definition 21

Let \( v_0, v_1, \ldots, v_n \) be vectors, and \( c_1, \ldots, c_n \) be constants which aren’t all just 0. Consider the equation
\[
c_1 v_0 + \cdots + c_n v_n = \vec{0}.
\]
If there are \( c_1, \ldots, c_n \) where this is true—except trivially when all are 0—then the vectors \( v_0, v_1, \ldots, v_n \) are said to be linearly dependent.

If this is always false—there are no \( c_1, \ldots, c_n \) except trivially when all are 0—then \( v_0, v_1, \ldots, v_n \) are said to be linearly independent.

Being linearly independent just means you can’t write one of the vectors in terms of the others by adding and scaling. Note that this is a property of collections of vectors, not the vectors themselves. \( \langle 1, 0 \rangle \) and \( \langle 0, 1 \rangle \) are linearly independent, but \( \langle 1, 0 \rangle, \langle 0, 1 \rangle, \langle 1, 1 \rangle \) are linearly dependent, since
\[
\langle 1, 1 \rangle = \langle 1, 0 \rangle + \langle 0, 1 \rangle.
\]

Testing linear independence can be done in a variety of ways. Note that the fundamental equation that we care about is
\[
c_1 v_0 + \cdots + c_n v_n = \vec{0}.
\]
If we regard the vectors \( v_i \) as column vectors, this is just the matrix equation
\[
\begin{bmatrix}
v_0 & \cdots & v_n
\end{bmatrix}
\begin{bmatrix}
c_0 \\
\vdots \\
c_n
\end{bmatrix}
= \vec{0}.
\]

On the left, each column is given by the \( v_i \) vectors, you might call each column the transpose of the vector.

V.3.A • 2. Result 24

Let \( A \) be a matrix. Thus \( \det(A) = 0 \) iff \( Ax = \vec{0} \) has a non-\( \vec{0} \) solution.

So to determine whether vectors are linearly independent, it’s enough to either calculate the determinant of the matrix of column vectors \( [v_0 \cdots v_n] \):
\[
\det[v_0 \cdots v_n] = 0 \iff \text{there is some } c_0, \ldots, c_n \text{ so that the above equality holds.}
\]
Alternatively, we can proceed by Gaussian elimination to attempt to find such \( c_0, \ldots, c_n \). If we end up only with \( c_0 = \cdots = c_n = 0 \), then we know that the vectors are linearly independent. Consider the following tests we then have.

V.3.A • 3. Result 25

A set of vectors \( \vec{v}_1, \vec{v}_2, \ldots, \vec{v}_n \) is linearly independent iff one of the three equivalent conditions holds:
\[
1. \det\begin{bmatrix}
  \vec{v}_1 \\
  \vdots \\
  \vec{v}_n
\end{bmatrix} \neq 0;
\]
\[
2. \det\begin{bmatrix}
  \vec{v}_1 & \cdots & \vec{v}_n
\end{bmatrix} \neq 0;
\]
\[
3. \begin{bmatrix}
  \vec{v}_1 & \cdots & \vec{v}_n
\end{bmatrix} \begin{bmatrix}
  \vec{c}_0 \\
  \vdots \\
  \vec{c}_n
\end{bmatrix}
= \vec{0}
\text{ only has the solution } \vec{c} = \vec{0}.
\]

We have the same sort of characterization for linear dependence.
Chapter V – Section 3.B

V.3.A • 4. Result 26

A set of vectors \(\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_n\) is linearly dependent iff one of the three equivalent conditions holds:

1. \(\det \begin{vmatrix} \vec{v}_1 \\ \vdots \\ \vec{v}_n \end{vmatrix} = 0;\)
2. \(\det \begin{vmatrix} \vec{v}_1 & \cdots & \vec{v}_n \end{vmatrix} = 0;\)
3. \(\begin{vmatrix} \vec{v}_1 & \cdots & \vec{v}_n \end{vmatrix} \vec{c} = \begin{vmatrix} 0 \\ \vdots \\ 0 \end{vmatrix}\) has a non-\(\vec{0}\) solution for \(\vec{c}\).

If we want an actual relationship between \(\vec{v}_1, \ldots, \vec{v}_n\), the solution of the third condition gives us one:

\(c_1 \vec{v}_1 + \cdots + c_n \vec{v}_n = \vec{0}\).

§ V.3.B. Eigenvalues and eigenvectors

An eigenvector is a vector which is only scaled by a given matrix. An eigenvalue is how much that vector is scaled by.

V.3.B • 1. Definition 22

Let \(A\) be a matrix. Consider the equation for \(\vec{v} \neq \vec{0}\)

\[A \cdot \vec{v} = \lambda \vec{v}.\]

\(\lambda\) is an eigenvalue for \(A\). \(\vec{v}\) is an eigenvector for \(A\) corresponding to the eigenvalue \(\lambda\).

Eigenvalues and eigenvectors will prove to be important for us later. Finding them can be a tricky process.

V.3.B • 2. Result 27

Let \(A\) be a square matrix. Thus \(\lambda\) is an eigenvalue iff

\(\det(A - \lambda \text{Id}) = 0.\)

\(\vec{v}\) is an eigenvector iff there is some eigenvalue \(\lambda_i\) where

\[(A - \lambda_i \text{Id}) \cdot \vec{v} = \vec{0}.\]

Proof:. . .

\(\lambda\) is an eigenvalue iff there is some non-\(\vec{0}\) \(\vec{v}\) where \(A \cdot \vec{v} = \lambda \vec{v}\). This is equivalent to

\(\vec{0} = A\vec{v} - \lambda \vec{v} = (A - \lambda \text{Id})\vec{v}.\)

By Result 24 (V.3.A • 2), this happens iff \(\det(A - \lambda \text{Id}) = 0.\)

\(\vec{v}\) is an eigenvector iff there is an eigenvalue where \(A = \lambda_i \vec{v}\). Moving to the other side and factoring yields that

this is equivalent to \((A - \lambda_i \text{Id})\vec{v} = \vec{0}.\)

So we just need to solve the polynomial equation \(\det(A - \lambda \text{Id}) = 0\) for various \(\lambda\). Now if we know all the eigenvalues, \(\lambda_1, \lambda_2, \ldots, \text{and } \lambda_n\), we can find the eigenvectors associated with each eigenvalue. Finding the eigenvectors is a simple matter of using Gaussian elimination to solve the equation

\[(A - \lambda_i \text{Id})\vec{v} = \vec{0}\]

for \(\vec{v}\), given the eigenvalue \(\lambda_i\). It’s important to note that with different eigenvalues we get different eigenvectors.\(^{\text{iii}}\)

So to get all the eigenvectors, we must consider all the eigenvalues, and solve this equation many times for differing \(\lambda_i\)s.

\(^{\text{iii}}\) to see this, suppose \(\vec{v}\) were an eigenvector associated to \(\lambda_1\) and \(\lambda_2\). Apply \(A\), and get that \(A\vec{v} = \lambda_1 \vec{v} = \lambda_2 \vec{v}\), which means \((\lambda_1 - \lambda_2)\vec{v} = \vec{0}.\) Since \(\vec{v} \neq \vec{0}\), and \(\lambda_1 - \lambda_2 \neq 0\), we get a contradiction, meaning there can be no such \(\vec{v}\).
Note that there will not be a unique solution, since if \( \mathbf{v} \) is an eigenvector, so is \( 2\mathbf{v} \), and \( \pi \mathbf{v} \), and any \( c\mathbf{v} \) for a constant \( c \neq 0 \). To see this, note that

\[
A(c\mathbf{v}) = cA\mathbf{v} = c\lambda \mathbf{v} = \lambda(c\mathbf{v})
\]

for some eigenvalue \( \lambda \).
Section V.4. Worked-Out Examples

V.4 • 1. Example 60

Consider the matrix equation
\[
\begin{bmatrix}
3 & -1 & -1 \\
-2 & 1 & -2 \\
4 & 2 & 3
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix}
= 
\begin{bmatrix}
-8 \\
-3 \\
30
\end{bmatrix}
\]

Use Gaussian elimination to solve for \(x, y, z\).

Solution ..:

We begin with the augmented matrix
\[
\begin{bmatrix}
3 & -1 & -1 & -8 \\
-2 & 1 & -2 & -3 \\
4 & 2 & 3 & 30
\end{bmatrix},
\]

and proceed to clear out the second column by adding the second row to the first, and subtracting twice the second row from the third:
\[
\begin{bmatrix}
1 & 0 & -3 & -11 \\
-2 & 1 & -2 & -3 \\
4 & 2 & 3 & 30
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & -3 & -11 \\
0 & 1 & -8 & -25 \\
0 & 0 & 31 & 124
\end{bmatrix}.
\]

From here, we can clear out the first column by adding the first row twice to the second, and \(-8\) times to the third.
\[
\begin{bmatrix}
1 & 0 & -3 & -11 \\
-2 & 1 & -2 & -3 \\
8 & 0 & 7 & 36
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & -3 & -11 \\
0 & 1 & -8 & -25 \\
0 & 0 & 31 & 124
\end{bmatrix}.
\]

Now we can scale the third row by \(1/31\) to get
\[
\begin{bmatrix}
1 & 0 & -3 & -11 \\
0 & 1 & -8 & -25 \\
0 & 0 & 1 & 4
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 0 & 1 \\
0 & 1 & 0 & 7 \\
0 & 0 & 1 & 4
\end{bmatrix}.
\]

and finally, we can clear out the third column:
\[
\begin{bmatrix}
1 & 0 & -3 & -11 \\
0 & 1 & -8 & -25 \\
0 & 0 & 1 & 4
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 0 & 1 \\
0 & 1 & 0 & 7 \\
0 & 0 & 1 & 4
\end{bmatrix}.
\]

This represents the matrix equation
\[
\begin{bmatrix}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix}
= 
\begin{bmatrix}
1 \\
7 \\
4
\end{bmatrix},
\]

i.e. \(x = 1, y = 7,\) and \(z = 4\).
V.4.2. Example 61

Solve the matrix equation

\[
\begin{bmatrix}
-6 & 12 & 21 \\
-1 & 5 & 3 \\
3 & -9 & -15
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix} =
\begin{bmatrix}
77 \\
18 \\
-52
\end{bmatrix}
\]

\[
\begin{align*}
R_2 &\rightarrow R_2 + 6R_1 \\
R_3 &\rightarrow 3R_2 \\
R_1 &\rightarrow R_1 + 15R_3 \\
R_3 &\rightarrow R_3 - 3R_1 \\
R_2 &\rightarrow R_2 + 5R_3 \\
\end{align*}
\]

\[
\begin{bmatrix}
1 & -5 & -3 & -18 \\
0 & -18 & 3 & -31 \\
0 & 6 & -6 & 2
\end{bmatrix}
\]

R_3 \rightarrow 3/6 R_3

\[
\begin{bmatrix}
1 & 0 & -23/6 & -169/18 \\
0 & 1 & -1/6 & 31/18 \\
0 & 0 & -5 & -25/3
\end{bmatrix}
\]

R_2 \rightarrow R_2 + 5/3 R_3

\[
\begin{bmatrix}
1 & 0 & 0 & -3 \\
0 & 1 & 0 & 2 \\
0 & 0 & 1 & 5/3
\end{bmatrix}
\]

\[
\begin{align*}
x &\rightarrow -3 \\
y &\rightarrow 2 \\
z &\rightarrow 5/3
\end{align*}
\]

V.4.3. Example 62

Solve the matrix equation

\[
\begin{bmatrix}
-2 & 4 & -3 & 1 \\
-3 & 2 & 2 & 3 \\
5 & 5 & 7 & -1 \\
-2 & -1 & 3 & 5
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z \\
t
\end{bmatrix} =
\begin{bmatrix}
-42 \\
-19 \\
33 \\
13
\end{bmatrix}
\]

\[
\begin{align*}
R_2 &\rightarrow R_2 + 3R_1 \\
R_3 &\rightarrow 5R_1, R_4 \rightarrow R_4 + 2R_1 \\
R_1 &\rightarrow R_1 + 3R_3 \\
\end{align*}
\]

\[
\begin{align*}
R_3 &\rightarrow 5R_1, R_4 \rightarrow R_4 + 2R_1 \\
R_1 &\rightarrow R_1 + 2R_3 \\
\end{align*}
\]

\[
\begin{align*}
R_2 &\rightarrow R_2 + 5R_3, R_4 \rightarrow R_4 - 3R_2 \\
\end{align*}
\]

\[
\begin{align*}
R_3 &\rightarrow R_3 + 5R_2, R_4 \rightarrow R_4 - 3R_3 \\
\end{align*}
\]

\[
\begin{align*}
R_2 &\rightarrow R_2 + 5R_3, R_4 \rightarrow R_4 - 3R_3 \\
\end{align*}
\]

\[
\begin{align*}
R_3 &\rightarrow R_3 + 5R_2, R_4 \rightarrow R_4 - 3R_3 \\
\end{align*}
\]

At this point, the standard way of proceeding would be to divide the third row by 252, and then cancel out the third column, and then do the same procedure for the fourth column. Doing this takes a lot of time, and it gets ugly. Still, the process is really just the same as before. If you’re feeling up to it, you can do this yourself, and check your result with the (surprisingly integer) answer \(\langle x, y, z, t \rangle = \langle 8, -5, 3, 3 \rangle\).
Find the inverse of the matrix

\[
\begin{bmatrix}
4 & 3 \\
2 & -6
\end{bmatrix}
\]

**Solution:**

We start out with the augmented matrix of the form

\[
\begin{bmatrix}
4 & 3 & 1 & 0 \\
2 & -6 & 0 & 1
\end{bmatrix}
\]

We perform the following operations:

\[
R_2 \rightarrow R_2 - 4R_1
\]

\[
\begin{bmatrix}
1 & -3 & 0 & 1/2 \\
0 & 15 & 1 & -2
\end{bmatrix}
\]

\[
R_2 \rightarrow 15R_2
\]

\[
\begin{bmatrix}
1 & -3 & 0 & 1 \\
0 & 1 & 1/15 & -2/15
\end{bmatrix}
\]

This is consistent with our general rule for calculating the inverse of 2 × 2 matrices:

\[
\begin{bmatrix}
4 & 3 \\
2 & -6
\end{bmatrix}^{-1} = \frac{1}{4 \cdot (-6) - 3 \cdot 2} \begin{bmatrix}
-3 & 2 \\
4 & -6
\end{bmatrix} = \frac{1}{30} \begin{bmatrix}
6 & 3 \\
2 & -4
\end{bmatrix} = \begin{bmatrix}
1/5 & 1/10 \\
1/15 & -2/15
\end{bmatrix}.
\]

Find the inverse of the matrix

\[
\begin{bmatrix}
-1 & 6 & 3 \\
2 & 2 & 4 \\
6 & -3 & 5
\end{bmatrix}
\]

**Solution:**

\[
\begin{bmatrix}
-1 & 6 & 3 & 1 & 0 & 0 \\
2 & 2 & 4 & 0 & 1 & 0 \\
6 & -3 & 5 & 0 & 0 & 1
\end{bmatrix}
\]

\[
R_1 \rightarrow -R_1
\]

\[
\begin{bmatrix}
1 & -6 & -3 & 1 & 0 & 0 \\
0 & 1 & 5/7 & 1/14 & 0 \\
0 & 0 & 4/7 & 9/7 & 33/14 & 1
\end{bmatrix}
\]

\[
R_2 \rightarrow 7R_3
\]

\[
\begin{bmatrix}
1 & 0 & 9/7 & -1/7 & 3/7 & 0 \\
0 & 1 & 5/7 & 1/7 & 1/14 & 0 \\
0 & 0 & 1 & -9/4 & 33/8 & -7/4
\end{bmatrix}
\]

\[
R_1 \rightarrow 9R_3
\]

\[
\begin{bmatrix}
1 & 0 & 0 & 11/4 & -39/8 & 9/4 \\
0 & 1 & 0 & -7/4 & -23/8 & 5/4 \\
0 & 0 & 1 & -9/4 & 33/8 & -7/4
\end{bmatrix}
\]
V.4 • 6. Example 65

Solve the system of equations, or show that there is no solution.

\[
\begin{align*}
x_1 + 2x_2 - x_3 &= 1 \\
2x_1 + x_2 + x_3 &= 1 \\
x_1 - x_2 + 2x_3 &= 1 
\end{align*}
\]

Solution :

Represent this by the matrix equation

\[
\begin{bmatrix}
1 & 2 & -1 \\
2 & 1 & 1 \\
1 & -1 & 2
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix} =
\begin{bmatrix}
1 \\
1 \\
1
\end{bmatrix}.
\]

Proceed by Gaussian elimination. Since the first entry is 1, we can clear out the rest of the column easily. Then we can scale.

\[
\begin{bmatrix}
1 & 2 & -1 \\
0 & -3 & 3 \\
0 & -3 & 3
\end{bmatrix}
\begin{bmatrix}
1 & 2 & -1 \\
0 & 1 & -1 \\
0 & -1 & 1
\end{bmatrix}
\begin{bmatrix}
1 & 2 & -1 \\
0 & 1 & -1 \\
0 & 0 & 0
\end{bmatrix}
\]

Thus the matrix equation requires 0z + 0y + 0x = 1/3, i.e. 0 = 1/3. Since this is false, there is no \(x, y, z\) where the hypotheses (the equations) are true. Hence there is no solution.

V.4 • 7. Example 66

Solve the system of equations, or show that there is no solution.

\[
\begin{align*}
x_1 + 2x_2 - x_3 &= 2 \\
2x_1 + x_2 + x_3 &= 1 \\
x_1 - x_2 + 2x_3 &= -1 
\end{align*}
\]

Solution :

Represent this by the matrix equation

\[
\begin{bmatrix}
1 & 2 & -1 \\
2 & 1 & 1 \\
1 & -1 & 2
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix} =
\begin{bmatrix}
2 \\
1 \\
-1
\end{bmatrix}.
\]

Proceed by Gaussian elimination. Since the first entry is 1, we can clear out the rest of the column easily. Then we can scale.

\[
\begin{bmatrix}
1 & 2 & -1 \\
0 & -3 & 3 \\
0 & -3 & 3
\end{bmatrix}
\begin{bmatrix}
1 & 2 & -1 \\
0 & 1 & -1 \\
0 & -1 & 1
\end{bmatrix}
\begin{bmatrix}
1 & 2 & -1 \\
0 & 1 & -1 \\
0 & 0 & 0
\end{bmatrix}
\]

This represents the equations

\[
\begin{align*}
x + z &= -1 \\
y - z &= 1
\end{align*}
\]

So there are many solutions depending on our choice of \(z\). In particular, \(x = -1 - z\), and \(y = 1 + z\). For example, \(z = 0\), and \(x = -1\) and \(y = 1\).
Determine whether the vectors
\[ x(t) = \langle e^{-t}, 2e^{-t} \rangle \quad y(t) = \langle e^{-t}, e^{-t} \rangle \quad z(t) = \langle 3e^{-t}, 0 \rangle. \]
are linearly independent for \(-\infty < t < \infty\)

**Solution:**

These are linearly dependent: \(3x - 6y + z = \langle 0, 0, 0 \rangle\). To see this, the matrix with columns \(x, y,\) and \(z\) can factor out an \(e^{-t}\):

\[
\begin{bmatrix}
e^{-t} & e^{-t} & 3e^{-t} \\
2e^{-t} & e^{-t} & 0 \
\end{bmatrix}
\begin{bmatrix}
a \\
b \\
c 
\end{bmatrix} =
\begin{bmatrix}
0 \\
0 
\end{bmatrix} \iff
e^{-t}
\begin{bmatrix}
1 & 1 & 3 \\
2 & 1 & 0 
\end{bmatrix}
\begin{bmatrix}
a \\
b \\
c 
\end{bmatrix} =
\begin{bmatrix}
0 \\
0 
\end{bmatrix}.
\]

Proceeding by Gaussian elimination yields the augmented matrix

\[
\begin{bmatrix}
1 & 0 & -3 & | & 0 \\
0 & 1 & 6 & | & 0 
\end{bmatrix}
\]

which states that \(a - 3c = 0,\) and \(b + 6c = 0.\) So any \(\langle a, b, c \rangle\) satisfying this witnesses that \(x, y,\) and \(z\) are linearly dependent. In particular, for any \(c,\) the following always works

\(\langle a, b, c \rangle = \langle 3c, -6c, c \rangle.\)

This yields the example above, where \(c = 1.\)

Find all the eigenvalues and eigenvectors of the matrix

\[
\begin{bmatrix}
5 & -1 \\
3 & 1 
\end{bmatrix}
\]

**Solution:**

Eigenvalues values \(\lambda\) where for some vector \(\vec{v} \neq 0,\)

\[A \cdot \vec{v} = \lambda \vec{v},\]

i.e. \((A - \lambda \text{Id}) \vec{v} = 0.\) Since \(\vec{v} \neq 0,\) this means \(A - \lambda \text{Id}\) isn’t invertible. This means \(\det(A - \lambda \text{Id}) = 0.\)

The reverse implication is the same sort of argument. But this means that the solutions to this equation are the eigenvalues.

\[
0 = \det \left( \begin{bmatrix} 5 & -1 \\ 3 & 1 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} \right)
\]

\[
= \det \left( \begin{bmatrix} 5 - \lambda & -1 \\ 3 & 1 - \lambda \end{bmatrix} \right)
\]

\[
= (5 - \lambda)(1 - \lambda) - 3(-1)
\]

\[
= 8 - 6\lambda + \lambda^2
\]

\[
= (\lambda - 4)(\lambda - 2).
\]

Thus the eigenvalues are 4 and 2. The associated eigenvectors can be found by solving the matrix equation

\[(A - \lambda \text{Id})\vec{v} = 0\]

for the vector \(\vec{v}\) for each eigenvalue \(\lambda.\)

\[
\lambda = 4 \quad \begin{bmatrix} 5 - 4 & -1 \\ 3 & 1 - 4 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}
\]
Find the eigenvalues and eigenvectors of the matrix

\[ A = \begin{bmatrix} -2 & 1 \\ 1 & -2 \end{bmatrix} \]

**Solution**:

The eigenvalues are the \( \lambda \) which satisfy

\[ \det(A - \lambda I) = 0. \]

We can calculate this as follows:

\[ 0 = \det(A - \lambda I) = \det\begin{bmatrix} -2 - \lambda & 1 \\ 1 & -2 - \lambda \end{bmatrix} = (-2 - \lambda)^2 - 1 \cdot 1 = \lambda^2 + 4\lambda + 3 \]

The quadratic formula tells us \( \lambda = -3 \) or \( \lambda = -1 \). So those are our eigenvalues. To find the associated eigenvectors, we solve \((A - \lambda I)v = 0\) for \(v\).

\[ \lambda = -3 \rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 0 \]

\[ x + y = 0 \quad \leftrightarrow \quad x = -y. \]

\[ \lambda = -1 \rightarrow \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 0 \]

\[ x - y = 0 \quad \leftrightarrow \quad x = y. \]

So we get eigenvector(s) \( \langle x, -x \rangle \) associated with the eigenvalue \( \lambda = -3 \), and eigenvector(s) \( \langle x, x \rangle \) associated with the eigenvalue \( \lambda = -1 \).

---

**Example 70**

Suppose that \( Ax = 0 \) for some vector \( x \neq 0 \). Show there is some \( y \neq 0 \) where \( A^* y = 0 \).

**Solution**:

The assumption tells us that \( \det(A) = 0 \). Now \( A^* y = 0 \) for some \( y \neq 0 \) iff \( \det(A^*) = 0 \). By definition, \( A^* = (\overline{A})^\dagger \). Yet the determinant of a matrix and its transpose is the same:

\[ \det(A^*) = \det((\overline{A})^\dagger) = \det(\overline{A}). \]

Because the determinant is given by sums and products of the entries, and conjugation distributes over these, \(^{iv}\) because \( (A - \lambda I)0 = 0 \), and \( (A - \lambda I)\overline{v} = 0 \). Applying the inverse of \( A - \lambda I \) to both sides would yield \( \overline{v} = 0 \), contradicting the hypothesis. So there can be no inverse.
i.e. \( \overrightarrow{a} + \overrightarrow{b} = \overrightarrow{a} + \overrightarrow{b} \), and \( \overrightarrow{a} \cdot \overrightarrow{b} = \overrightarrow{a} \cdot \overrightarrow{b} \), it follows that \( \text{det}( \overrightarrow{A}^* ) = \text{det}( \overrightarrow{A} ) = 0. \)

Hence \( \text{det}( \overrightarrow{A}^* ) = 0 \), giving the result.

### V.4 • 12. Example 71

Solve the system of equations

\[
\begin{align*}
x - 2y + z + u &= 1 \\
2x + y - z - 2u &= 2 \\
3x + 3y + 0z + 2u &= 3 \\
x + y + z + u &= 4
\end{align*}
\]

by Gaussian elimination.

**Solution:**

We first should convert this into a matrix equation:

\[
\begin{bmatrix}
1 & -2 & 1 & 1 \\
2 & 1 & -1 & -2 \\
3 & 3 & 0 & 2 \\
1 & 1 & 1 & 1
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z \\
u
\end{bmatrix}
= 
\begin{bmatrix}
1 \\
2 \\
3 \\
4
\end{bmatrix}
\]

Now we form the augmented matrix, and proceed by Gaussian elimination.

\[
\begin{bmatrix}
1 & -2 & 1 & 1 & 1 \\
2 & 1 & -1 & -2 & 0 \\
3 & 3 & 0 & 2 & 0 \\
1 & 1 & 1 & 1 & 0
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 7 & 9 & 13 \\
0 & 1 & 3 & 4 & 6 \\
0 & 0 & -9 & -12 & -15 \\
0 & 0 & 0 & 0 & 3
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 7 & 9 & 13 \\
0 & 1 & 3 & 4 & 6 \\
0 & 0 & 1 & 4/3 & 5/3 \\
0 & 0 & 0 & 1 & 4/3
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 0 & -1/3 & 4/3 \\
0 & 1 & 0 & 0 & 1 \\
0 & 0 & 1 & 4/3 & 5/3 \\
0 & 0 & 0 & 1 & -4/3
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & 0 & 0 & 8/9 \\
0 & 1 & 0 & 0 & 1 \\
0 & 0 & 1 & 0 & 31/9 \\
0 & 0 & 0 & 1 & -4/3
\end{bmatrix}
\]

This represents the equation

\[
\begin{bmatrix}
8/9 \\
1 \\
31/9 \\
-4/3
\end{bmatrix}
= 
\begin{bmatrix}
1 & 0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 & 1 \\
0 & 0 & 1 & 0 & 1 \\
0 & 0 & 0 & 1 & 0
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z \\
u
\end{bmatrix}
= 
\begin{bmatrix}
x \\
y \\
z \\
u
\end{bmatrix}
\]

so that our solution \( \langle x, y, z, u \rangle \) is just \( \langle 8/9, 1, 31/9, -4/3 \rangle \).
1. Solve by Gaussian elimination the system

\[
\begin{align*}
  x + 2y - z + u &= 0 \\
  2x - y + 3z + 2u &= 2 \\
  x - 3y + z - u &= 3 \\
  -x + y - z + u &= a.
\end{align*}
\]

2. For what value of \( a \) does \( u = 0 \)?

**Solution:**

1. Proceed by Gaussian elimination as follows:

\[
\begin{bmatrix}
  1 & 2 & -1 & 1 & 0 \\
  2 & -1 & 3 & 2 & 0 \\
  1 & -3 & 1 & -1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix}
  1 & 2 & -1 & 1 & 0 \\
  0 & -5 & 5 & 0 & 2 \\
  -1 & 1 & -1 & 1 & a \end{bmatrix} \\
\begin{bmatrix}
  1 & 2 & -1 & 1 & 0 \\
  0 & -5 & 2 & -2 & 3 \\
  0 & 3 & -2 & 2 & a \end{bmatrix} \rightarrow \begin{bmatrix}
  1 & 2 & -1 & 1 & 0 \\
  0 & 1 & -1 & 0 & -2/5 \\
  0 & 3 & -2 & 2 & a \end{bmatrix} \\
\begin{bmatrix}
  1 & 0 & 1 & 1 & 4/5 \\
  0 & 1 & -1 & 0 & -2/5 \\
  0 & 0 & 1 & 2 & a + 6/5 \end{bmatrix} \rightarrow \begin{bmatrix}
  1 & 0 & 1 & 1 & 1 \\
  0 & 1 & -1 & 0 & 0 \\
  0 & 0 & 1 & 2 & 0 \end{bmatrix} \\
\begin{bmatrix}
  1 & 0 & 0 & -1 & a - 2/5 \\
  0 & 1 & 0 & 2 & a + 4/5 \\
  0 & 0 & 1 & 2 & a + 6/5 \end{bmatrix} \rightarrow \begin{bmatrix}
  1 & 0 & 0 & 0 & -a/4 + 3/4 \\
  0 & 1 & 0 & 2 & -a/2 - 3/2 \\
  0 & 0 & 1 & 2 & -a/2 - 11/10 \end{bmatrix} \\
\begin{bmatrix}
  1 & 0 & 0 & 0 & 3a/4 + 23/20 \\
  0 & 1 & 0 & 2 & 3a/4 + 23/20 \\
  0 & 0 & 1 & 2 & 3a/4 + 23/20 \end{bmatrix}
\]

Hence we can say that

\[
\begin{align*}
  x &= \frac{3 - a}{4}, \\
  y &= \frac{-a - 3}{2}, \\
  z &= \frac{-5a - 11}{10}, \\
  u &= \frac{15a + 23}{20}.
\end{align*}
\]

2. \( u = 0 \), by the result above, is equivalent to having \( 15a + 23 = 0 \), i.e. \( a = -23/15 \).
CHAPTER V – SECTION 4

82
Chapter VI. Linear Systems of Differential Equations

We will look at differential equations which make statements about multiple functions. Most of the time, these functions will interact in some way, in the sense that the equalities depend on their values together.

In particular, we have the following definition for what we mean by a “system of differential equations”.

VI • 1. Definition 23

A system of differential equations is a sequence of equations of the form
\[
\begin{align*}
x_1' &= F_1(x_1, \cdots, x_n, t) \\
x_2' &= F_2(x_1, \cdots, x_n, t) \\
&\vdots \\
x_n' &= F_n(x_1, \cdots, x_n, t).
\end{align*}
\]

A solution to such a system is a sequence of functions \(x_1(t), x_2(t), \cdots, x_n(t)\) which satisfy those equations.

We will be interested in equations that are linear in form, as well as those which are homogeneous. These definitions are analogous to those in a single variable, and with the language of matrices, these definitions become practically the same. This is because a system of linear differential equations will be a sequence of equations
\[
\begin{align*}
x_1' &= p_{11}(t)x_1 + \cdots + p_{1n}(t)x_n + g_1(t) \\
x_2' &= p_{21}(t)x_1 + \cdots + p_{2n}(t)x_n + g_2(t) \\
&\vdots \\
x_n' &= p_{n1}(t)x_1 + \cdots + p_{nn}(t)x_n + g_n(t).
\end{align*}
\]

This can then be represented by the matrix equation
\[
\begin{bmatrix}
x_1' \\
\vdots \\
x_n'
\end{bmatrix} =
\begin{bmatrix}
p_{11}(t) & \cdots & p_{1n}(t) \\
\vdots & \ddots & \vdots \\
p_{n1}(t) & \cdots & p_{nn}(t)
\end{bmatrix}
\begin{bmatrix}
x_1 \\
\vdots \\
x_n
\end{bmatrix} +
\begin{bmatrix}
g_1(t) \\
\vdots \\
g_n(t)
\end{bmatrix},
\]

for various functions \(p_{ij}\) and \(g_i\) where \(i, j \leq n\). In other words,
\[
z' = A(t) \cdot z + G(t),
\]

for matrices of functions \(A(t)\) and \(G(t)\). The derivative \(z'\) is just component-wise: we differentiate each entry in \(z\).

VI • 2. Definition 24

A system of differential equations is linear iff it is of the form
\[
z' = A(t) \cdot z + G(t),
\]

for matrices of functions \(A(t)\) and \(G(t)\), and vector of variables \(z\).

A system of differential equations is homogeneous iff \(G(t)\) is constantly 0 above: it’s of the form
\[
z' = A(t) \cdot z.
\]

So in the case that \(A\) is a constant matrix—i.e. just filled with constants—this situation simplifies considerably, as we’ll see later.

For now, these equations can be useful for thinking about higher order differential equations, since we can go back and
Chapter VI – Section 1

forth between the two for systems of linear equations.

<table>
<thead>
<tr>
<th>VI • 3. Result 28</th>
</tr>
</thead>
<tbody>
<tr>
<td>Every linear ( n )-th order differential equation can be represented as a system of linear differential equations of ( n ) variables, and the reverse holds under certain conditions.</td>
</tr>
</tbody>
</table>

**Proof**: For one direction, consider the linear \( n \)-th order differential equation

\[
p_n(t)x^{(n)} + \cdots + p_1(t)x' + p_0(t)x = g(t).
\]

The system of differential equations

\[
\begin{align*}
x &= x_0 \\
x' &= x_1 \\
P_0 &= x_2 \\
&\vdots \\
x_{n-2} &= x_{n-1} \\
p_n(t)x_{n-1} + p_{n-1}(t)x_{n-2} + \cdots + p_1(t)x_1 + p_0(t)x_0 &= g(t)
\end{align*}
\]

then represents the \( n \)-th order differential equation, since \( x_i \) just represents the \( i \)th derivative of \( x \).

The other direction is a lengthy process of differentiating equations, solving for a variable to reduce to fewer variables, and so on until we finish. So this will be a proof by example in the 2 × 2 case in Appendix A. ⊢

**Section VI.1. Homogeneous Systems with Constant Coefficients**

When dealing with systems of differential equations, a base case to consider is when the equations are linear, meaning of the form

\[
z' = A(t) \cdot z + G(t)
\]

for matrices \( A \) and \( G \) depending on \( t \). To simplify this situation, we will consider for now equations where these are constant, and where \( G = 0 \). In other words, the equation

\[
z' = A \cdot z
\]

for some (constant) matrix \( A \). To simplify the situation even further, we’ll work just with \( 2 \times 2 \) systems for now.

<table>
<thead>
<tr>
<th>VI.1 • 1. Result 29</th>
</tr>
</thead>
</table>
| Let \( A \) be a \( 2 \times 2 \) matrix, and \( z(t) \) a vector. Consider the differential equation

\[
z' = A \cdot z
\]

Suppose \( A \) has distinct eigenvalues \( \lambda_1 \) and \( \lambda_2 \) with some associated eigenvectors \( v \) and \( w \) respectively. Thus solutions to the above differential equation are of the form

\[
z(t) = Cve^{\lambda_1 t} + Dwe^{\lambda_2 t},
\]

for some constants \( C \) and \( D \). |

**Proof**: The proof of this can be found in Appendix A

This generalizes to complex valued eigenvectors and eigenvalues, but often we want to work only with real-valued
functions. In this case, we can simplify the answer using Euler’s formula, and choosing different constants.

**VI.1 • 2. Result 30**

Let $A$ be a $2 \times 2$ matrix with real entries, and $z(t)$ a vector. Consider the differential equation

$$z' = A \cdot z.$$ 

Suppose $A$ has complex eigenvalue $a + ib$ ($b \neq 0$) and associated eigenvector $\tilde{v} + i\tilde{w}$.

Thus the eigenvectors of $A$ are $a \pm ib$ with eigenvectors $\tilde{v} \pm i\tilde{w}$, and solutions to the differential equation are

$$z(t) = Ce^{at}((\tilde{v} \cos(bt) - \tilde{w} \sin(bt)) + De^{at}(\tilde{v} \sin(bt) + \tilde{w} \cos(bt))),$$

for some constants $C$ and $D$.

Unfortunately, this is the simplest we can do. In the single variable case we get $Ce^{at} \sin(bt) + De^{at} \cos(bt)$, which corresponds to this under certain choices of $C$, $D$, and scaling $\tilde{v}$ and $\tilde{w}$. Unfortunately, this just makes the other entry in $z$ uglier. So to be fair, both are given an equal amount of ugliness. In practice, however, the ugliness will often be skewed towards one solution with the other being merely $Ce^{at} \sin(bt) + De^{at} \cos(bt)$.

Just like with a single variable, if there is only one eigenvalue—or solution to the characteristic equation in the case of a single variable—the solution isn’t necessarily like in Result 29 (VI.1 • 1). And the result isn’t so simply stated: the solution actually depends on whether the eigenvectors are linearly independent, and we get a somewhat different form from Result 29 (VI.1 • 1) if they are linearly dependent.

**VI.1 • 3. Result 31 (2×2 Systems)**

Let $A$ be a $2 \times 2$ matrix, and $z(t)$ a vector. Consider the differential equation

$$z' = Az.$$ 

Suppose $A$ has eigenvalue $\lambda_1$ with eigenvector $v_1 \neq \bar{0}$, and eigenvalue $\lambda_2$ with eigenvector $v_2 \neq \bar{0}$.

1. If $\lambda_1 \neq \lambda_2$, then

$$z(t) = Cv_1 e^{\lambda_1 t} + Dv_2 e^{\lambda_2 t}.$$ 

2. If $\lambda_1 = \lambda_2 = \lambda$, but $v_1$ and $v_2$ are linearly independent, then

$$z(t) = Cv_1 e^{\lambda t} + Dv_2 e^{\lambda t}.$$ 

3. If $\lambda_1 = \lambda_2 = \lambda$, and $v_1$ and $v_2$ are linearly dependent, then

$$z(t) = Cv_1 e^{\lambda t} + D(v_2 \cdot t \cdot e^{\lambda t} + me^{\lambda t}),$$

where $m$ is a vector solving the matrix equation $(A - \lambda \text{Id})m = v_2$.

Where $C$ and $D$ are constants.

Note that the choice of $v_2$ in the equation of $m$ in case (3) was arbitrary. Because $v_1$ and $v_2$ were linearly dependent, they were a constant multiple of each other: $v_1 = kv_2$ for some $k \neq 0$. In this case, we could just as well say that $v$ is the only eigenvector (up to scaling), and write

$$z(t) = Cve^{\lambda t} + D(v \cdot t \cdot e^{\lambda t} + me^{\lambda t})$$

for $m$ solving $(A - \lambda \text{Id})m = v$. This really just results in absorbing the constant into $D$. The reason why such an $m$ pops up is that we need a fundamental set of solutions, just like in the single variable case.

Now as a process of solving such systems, **2×2 Systems** (VI.1 • 3) translates to the series of steps:

1. Find the eigenvalues of $A$.
2. Find the corresponding eigenvectors of $A$.
3. Determine if the eigenvectors are linearly independent or linearly dependent.

---

1 In particular, if we fix some row, we’re scaling $\tilde{v} + i\tilde{w}$ to make the entry in that row $1$, and then dealing with the new eigenvectors where the complex values only occur in the other row.
4. If linearly dependent solve \((A - \lambda I)d = v\) for an eigenvector \(v\).
5. Use the appropriate solution as in 2 × 2 Systems (VI.1 • 3).

More concretely, we get the series of steps on how to do the above.

1. Solve \(\det(A - \lambda I) = 0\) for \(\lambda\) by solving a polynomial equation.
2. For each \(\lambda\) above, solve \((A - \lambda I)v = 0\) for \(v\) using Gaussian elimination.
   - If \(\lambda_1 \neq \lambda_2\), then \(z(t) = C_1 e^{\lambda_1 t} + D_1 e^{\lambda_2 t}\).
   - If \(\lambda_1 = \lambda_2 = \lambda\), but \(v_1\) and \(v_2\) aren’t multiples of each other, then \(z(t) = C_1 e^{\lambda t} + D_2 e^{\lambda t}\).
3. If there is only one \(v\) up to scaling, then solve \((A - \lambda I)m = v\) for \(m\) using Gaussian elimination.
   - In this case, \(z(t) = C_1 e^{\lambda t} + D(v \cdot t e^{\lambda t} + m e^{\lambda t})\).

Note that having one eigenvalue with linearly independent eigenvectors is pretty rare. In fact, for the 2 × 2 case, this happens only when \(A\) is a multiple of the identity matrix \(I\).

Let’s continue with 2 × 2 systems for a moment to deal with complex values.

<table>
<thead>
<tr>
<th>VI.1 • 4. Result 32 (Complex 2 × 2 Solutions)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Let (A) be a 2 × 2 matrix with real entries, and (z(t)) a vector. Consider the differential equation (z' = A \cdot z). Suppose (A) has complex eigenvalue (a + ib) ((b \neq 0)) and associated eigenvector (v + iw). Thus the eigenvectors of (A) are (a \pm ib) with eigenvectors (v \pm iw), and solutions to the differential equation are (z(t) = Ce^{at}(v \cos(bt) - w \sin(bt)) + De^{at}(v \sin(bt) + w \cos(bt))), for some constants (C) and (D).</td>
</tr>
</tbody>
</table>

Showing all of these facts can be done just by converting them into homogeneous equations with constant coefficients, each of a single variable, and then solving them in the usual way we do there. In the case of complex eigenvalues, we merely use Euler’s formula, and change the constants to remove any \(i\).

Let’s generalize this idea to \(n \times n\) systems. Most larger systems will still be fairly small. Even the small 3 × 3 is still computationally difficult to deal with. But the ideas from the 2 × 2 systems generalize to \(n \times n\) systems easily in the case that the eigenvalues are distinct.

<table>
<thead>
<tr>
<th>VI.1 • 5. Result 33 (n × n systems)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Let (A) be an (n \times n) matrix, and (z(t)) a vector. Consider the differential equation (z' = Az). Suppose (A) has eigenvalues (\lambda_1, \lambda_2, \ldots, \lambda_n)—which aren’t necessarily different—with corresponding eigenvectors (v_1, v_2, \ldots, v_n). If the eigenvectors (v_1, \ldots, v_n) are linearly independent or the eigenvalues (\lambda_1, \ldots, \lambda_n) are all distinct, then (z(t) = C_1 v_1 e^{\lambda_1 t} + C_2 v_2 e^{\lambda_2 t} + \cdots + C_n v_n e^{\lambda_n t}), for constants (C_1, \cdots, C_n).</td>
</tr>
</tbody>
</table>

\(^{ii}\) To see this, otherwise \(A - \lambda I\) isn’t the \(0\) matrix. And thus we have an equation with at least one unknown: \((A - \lambda I)v = 0\). But then either one variable is fixed (e.g. if one equation only involves one unknown), or \(a\) and \(b\) have some sort of scalar relationship, which tells us that the eigenvectors can’t be linearly independent.
If we have an eigenvalue $\lambda$ repeated twice with only one (linearly independent) eigenvector, then instead of contributing the term $Cve^{\lambda t} + Dve^{\lambda t}$ to $z(t)$, we add to $z(t)$

$$Cve^{\lambda t} + D(tve^{\lambda t} + m),$$

where $(A - \lambda I)^2m = 0 \neq m$.

If you go on to study more linear algebra, $m$ is called a generalized eigenvector in that $(A - \lambda I)^k m = 0$ for some $k$.

In the case that $k = 1$, satisfying this just means being eigenvector associated to $\lambda$—so long as the vector isn’t $0$.

To continue this aside, things get more complicated as more and more eigenvalues are repeated. If an eigenvalue $\lambda$ is repeated $n$ times with one eigenvector $v$ (up to scaling), then instead of adding $Cve^{\lambda t}$ $n$ times—yielding only one term when we need $m$—we add instead

$$Cte^{\lambda t} \sum_{k=1}^{n} \left( m_k \frac{t^k}{k!} \right),$$

for each $i \leq n$. Each $m_k$ is defined by recursion:

$$m_0 := v_i$$
$$m_1 := \text{a solution to } (A - \lambda I)m = m_0$$
$$\vdots$$
$$m_{k+1} := \text{a solution to } (A - \lambda I)m = m_k.$$  

So if we’re working in a $3 \times 3$ system, and we have eigenvalues $\lambda_1$ and $\lambda_2 = \lambda_3$, having eigenvectors $v_1$ and $v_2$ respectively, then the general solution will be

$$z(t) = Cv_1e^{\lambda_1 t} + Dv_2e^{\lambda_2 t} + (tv_2 + m).$$

If we have just one eigenvalue $\lambda = \lambda_1 + \lambda_2 + \lambda_3$ with just one eigenvector $v$, then the solution will be

$$z(t) = C_1ve^{\lambda t} + C_2ve^{\lambda t}(tv + m_1) + C_3e^{\lambda t}\left( \frac{t^2}{2} + tm_1 + m_2 \right).$$

In general, however, it’s simpler to refer to notions of “exponentiation” with matrices defined along the lines of a power series used in the case of real numbers. In this case, $z = A\bar{z}$ has a solution of $\exp(At)\bar{c}$ for a (column) vector of constants $\bar{c}$.

**Section VI.2. Non-Homogeneous Systems**

Many of the results about linear, differential equations—of a single variable—apply also to systems of such equations. In particular, the general solution of a homogeneous $n \times n$ system is one where we have $n$ “different enough” solutions $\phi_1$, through $\phi_n$, in that they have a non-zero Wronskian matrix. The general solution in that case is just $C_1\phi_1 + \cdots + C_n\phi_n$. If we have a non-homogeneous system, the difference is still a single, particular solution:

**VI.2.1. Result 34**

Consider the $n \times n$ linear system of differential equations

$$z' = A(t) \cdot z + G(t).$$

Let $Z(t)$ be any solution to this. Let $\phi_1, \phi_2, \cdots, \phi_n$ solve the homogeneous version: $\phi' = A(t) \cdot \phi$.

Thus the general solution is

$$z(t) = C_1\phi_1(t) + C_2\phi_2(t) + \cdots + C_n\phi_n(t) + Z(t)$$

for constants $C_1, \ldots, C_n$. 

87
Consider an arbitrary solution $z$ of the non-homogeneous differential equation. Consider the function $\Phi = z - Z$.

Note that then

\[
\Phi' = z' - Z' = A(t) \cdot z + G(t) - (A(t) \cdot Z + G(t)) = A(t) \cdot (z - Z) = A(t) \cdot \Phi.
\]

Hence $\Phi$ is a solution to the homogeneous version. Hence $\Phi = z - Z$ is of the form $\Phi = C_1 \phi_1 + \cdots + C_n \phi_n$ for constants $C_1, \ldots, C_n$. But then, adding $Z$ to both sides, $z = C_1 \phi_1 + \cdots + C_n \phi_n + Z$.

So to solve non-homogeneous equations, we have similar methods as in the single variable case. In particular, what follows is the $n \times n$ version of undetermined coefficients.

### VI.2.2. Result 35

Consider the $n \times n$ linear system of differential equations $z' = A \cdot z + G(t)$.

- If $g(t) = \tilde{c}_n t^n + \cdots + \tilde{c}_1 t + \tilde{c}$, then guess $Y(t) = \tilde{C}_n t^n + \cdots + \tilde{C}_1 t + \tilde{C}_0$;
- If $g(t) = \tilde{c} \sin(at)$, or $g(t) = \tilde{c} \cos(at)$, then guess $Y(t) = \tilde{C} \sin(at) + \tilde{D} \cos(at)$;
- If $g(t) = \tilde{c} e^{at}$, then guess $Y(t) = \tilde{C} e^{at}$;

If this guess doesn’t work—i.e. it’s a solution to the homogeneous version—then keep multiplying the guess by $t$—including lower degrees—until it doesn’t.

For example, you might initially guess $\tilde{c} e^{at}$. If that doesn’t work, you would consider not just $\tilde{c} t e^{at}$, but $\tilde{c} t e^{at} + \tilde{d} e^{at}$. If that doesn’t work, you would then consider $\tilde{c}_1 t^2 e^{at} + \tilde{c}_2 t e^{at} + \tilde{c}_3 e^{at}$. And so on and so forth until it does work.

Note also that, just like the single variable case, undetermined coefficients can only be used in the case that the linear system has constant coefficients.

### Section VI.3. Linearization

There are many more kinds of differential equations beyond just ones of the form

\[
z' = A(t)z + G(t),
\]

for matrix functions $A$ and $G$. We really only know, in principle, how to solve them when $A$ is constant, and $G$ is nice enough. But more generally, systems of differential equations are of the form

\[
z' = F(t, z),
\]

with no real restrictions on $F^{iii}$. Let’s consider the differential equation of this form

\[
z' = \begin{bmatrix} F_1'(z) \\ F_2'(z) \\ \vdots \\ F_n'(z) \end{bmatrix},
\]

where we at least have that each $F_i$ is differentiable and doesn’t depend on $t$. We can approximate such a solution using an approximation to the differential equation. To do this, assume $F(s) = 0$ at some (vector) value $s$. Now we use a linear approximation to $F$ at $s$, and consider the new, linear system of differential equations around $s$.

\[iii\text{beyond the fact that it outputs a vector the size of } z.\]
That’s a lot to take in, so let’s just consider the $2 \times 2$ case:

\[
\begin{pmatrix}
x' \\
y'
\end{pmatrix} = \begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix}
\]

for differentiable functions $f$ and $g$. Suppose $f(x_1, y_1) = g(x_1, y_1) = 0$ for numbers $x_1$ and $y_1$. Such a point $(x_1, y_1)$ is a critical point for $(f, g)$. Because of this, we have approximations to $f(x, y)$ and $g(x, y)$ near $s = (x_1, y_1)$ by a Taylor approximation: denoting $\frac{\partial f}{\partial x}$ by $f_x$ and so forth,

\[
f(x, y) \approx f(s) + f_x(s)(x - x_1) + f_y(s)(y - y_1)
\]

\[
g(x, y) \approx g(s) + g_x(s)(x - x_1) + g_y(s)(y - y_1)
\]

 Hence we can write a different system of equations whose solution is not the same as before, but which is supposed to approximate the solution to $z' = F(z)$ when $z$ is near $s$: choose new variables $u = x - x_1$, $v = y - y_1$, and consider the new system

\[
\begin{pmatrix} u' \\ v' \end{pmatrix} \approx \begin{pmatrix} f_x(s) & f_y(s) \\ g_x(s) & g_y(s) \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix}.
\]

the system above—replacing $\approx$ with $\equiv$—is called the linearization of the original system. The variables $u$ and $v$ are approximations to $x - x_1$ and $y - y_1$ near 0. The matrix on the left is the Jacobian of $F$, denoted $J(F)$, evaluated at $s$.

### VI.3.1. Definition 25

Consider the differential equation

\[
z' = \begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix} = F(z)
\]

for continuously differentiable functions $f$ and $g$. Suppose $F(s) = 0$ for vector $s = (x_1, y_1)$, a critical point of $F$.

The linearization of $z' = F(z)$ is the linear system

\[
w' = J(F)(s) \cdot w = \begin{pmatrix} f_x(x_1, y_1) & f_y(x_1, y_1) \\ g_x(x_1, y_1) & g_y(x_1, y_1) \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix},
\]

where $(u, v) = w \approx z - s = (x - x_1, y - y_1)$.

Just like with autonomous differential equations—equations of the form $y' = f(y)$—there is a notion of stability for critical points of $z' = F(z)$. This notion is slightly complicated by the fact that we have many functions to worry about, but ultimately, the concept is similar. Consider first the simple case of when we’re dealing with homogeneous equations.

So long as $\det(A) \neq 0$, the only critical point of $z' = Az$ is $z = 0$. This point is asymptotically stable if solutions $z$ go to this critical point $0$ as $t \to \infty$. We know that solutions to such equations are usually of the form

\[
C_1v_1e^{\lambda_1t} + \cdots + C_nv_ne^{\lambda_nt}.
\]

Given that all those $C_i v_i$ are all just constant vectors, the real determiner of where the solutions go is all those $e^{\lambda_i}$s, and so the eigenvalues. If all the eigenvalues have a real part that’s negative, then such solutions go to $0$.

So the notion of stability is roughly the same, and it is contingent on whether functions that make up the solution go to 0, or go to $\infty$. This is really determined by the eigenvalues of the matrix $J(F)(s)$. 


VI.3 • 2. Definition 26

Consider the differential equation \( \dot{z} = Az \). Let \( \lambda_1 \) and \( \lambda_2 \) be eigenvalues for \( A \).

1. The critical point \( \vec{0} \) is asymptotically stable iff \( e^{\lambda_1 t} \) and \( e^{\lambda_2 t} \) both go to 0 as \( t \to \infty \). In other words, \( \lambda_1 \) and \( \lambda_2 \) (or at least their real part) are negative.

2. The critical point \( \vec{0} \) is stable iff \( e^{\lambda_1 t} \) and \( e^{\lambda_2 t} \) are each constant or have no limit (not \( \infty \), and not 0) as \( t \to \infty \). In other words, \( \lambda_1 \) and \( \lambda_2 \) (or at least their real part) are 0.

3. The critical point \( \vec{0} \) is unstable iff \( e^{\lambda_1 t} \) or \( e^{\lambda_2 t} \) goes to \( \infty \) as \( t \to \infty \). In other words, \( \lambda_1 \) or \( \lambda_2 \) (or at least their real part) is positive.

If we look back at Definition 25 (VI.3 • 1), if \( w = \vec{0} \) is an asymptotically stable critical point of the linearization, then \( w \approx z - s \) goes to \( \vec{0} \) as \( t \to \infty \), i.e. \( z \) converges to \( s \). Similarly, if \( w \) goes to \( \infty \), then \( z - s \) does not go to 0, and so \( z \) diverges from \( s \). This is just to say that by looking at the linearization, we can tell whether solutions near \( z \) converge to \( s \) or not.

VI.3 • 3. Result 36

Consider the differential equation \( \dot{z} = F(z) \) with linearization \( \dot{w} = J(F)(s) \cdot w \) around the critical point \( s \). Therefore,

1. the critical point \( z = s \) is asymptotically stable if \( w = \vec{0} \) is; and

2. the critical point \( z = s \) is unstable if \( w = \vec{0} \) is.

Note that this is really all determined by our choice of \( s \). We look at the linearization \( (z - s)' = A(z - s) \), and see whether \( z - s = \vec{0} \) is an asymptotically stable or unstable critical point, and thus whether solutions \( z \) near \( s \) go to \( s \) or not.
Section VI.4. Worked-Out Examples

VI.4 • 1. Example 73

Transform the given initial value problem into an initial value problem for two first-order equations.
\[ u'' + p(t)u' + q(t)u = g(t), \quad u(0) = u_0, \quad u'(0) = u'_0. \]

**Solution:**
With initial conditions \( u_1(0) = u'_0, \) and \( u_0(0) = u_0, \)
\[ u'_1 = u_0, \quad u''_1 = g - pu_1 - qu_0. \]

VI.4 • 2. Example 74

Consider the system
\[
\begin{align*}
\dot{x}_1 &= -2x_1 + x_2 \\
\dot{x}_2 &= x_1 - 2x_2.
\end{align*}
\]

a. Solve the first differential equation for \( x_2. \)
b. Substitute the result of (a) into the second differential equation, thereby obtaining a second-order differential equation for \( x_1. \)
c. Solve the differential equation in (b) for \( x_1. \)
d. Use (a) and (c) to find \( x_2. \)

**Solution:**
\[
\begin{align*}
\dot{x}_2 &= x_1 - 2x_2 \iff x''_2 + 2x'_2 = x_1 - 2(x'_1 + 2x_1) \\
&\iff x''_1 + 4x'_1 + 3x_1 = 0.
\end{align*}
\]

This is a homogeneous, second-order, differential equation with constant coefficients. Solutions of the form \( e^{rt} \) occur for
\[ r^2 + 4r + 3 = 0 \iff r = -3 \text{ or } r = -1. \]
Hence the general form of the solution for \( x_1 \) is
\[ x_1(t) = Ce^{-3t} + De^{-t}. \]
for constants \( C \) and \( D. \)
d. \( x'_1(t) = -3Ce^{-3t} - De^{-t} \) so the equality \( x_2 = x'_1 + 2x_1 \) yields
\[ x_2(t) = -Ce^{-3t} + De^{-t}. \]

§ VI.4.A. 2 × 2 Systems

VI.4.A • 1. Example 75

Solve the differential equation
\[ \dot{z} = Az = \begin{bmatrix} 3 & 2 \\ -1 & 6 \end{bmatrix} z \]
for the vector function \( z(t). \)

**Solution:**
The first step here is to identify the eigenvalues. To do this, we solve \( \det(A - \lambda \text{Id}) = 0 \) for \( \lambda \).

\[
\det(A - \lambda \text{Id}) = 0 \quad \iff \quad \det \begin{bmatrix} 3 - \lambda & 2 \\ -1 & 6 - \lambda \end{bmatrix} = 0 \\
\iff (3 - \lambda)(6 - \lambda) + 2 = 0 \\
\iff 20 - 9\lambda + \lambda^2 = 0 \\
\iff \lambda = 4 \text{, or } \lambda = 5.
\]

So we have eigenvalues 4 and 5. Now we must find some associated eigenvectors. To do this, we consider \((A - \lambda \text{Id})v = 0\) for each eigenvalue \( \lambda \), and solve for \( v \) in each case. This is accomplished by Gaussian elimination.

\[
\lambda = 4 \quad \iff \quad \begin{bmatrix} 3 - 4 & 2 \\ -1 & 6 - 4 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
\iff \begin{bmatrix} -1 & 2 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
\iff -v_1 + 2v_2 = 0 \\
\iff v_1 = 2v_2.
\]

\[
\lambda = 5 \quad \iff \quad \begin{bmatrix} 3 - 5 & 2 \\ -1 & 6 - 5 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
\iff \begin{bmatrix} -2 & 2 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
\iff -v_1 + v_2 = 0 \\
\iff v_1 = v_2.
\]

Hence we get eigenvectors \( \langle 2, 1 \rangle \) with eigenvalue 4 and \( \langle 1, 1 \rangle \) with eigenvalue 5 (as well as any non-zero scaling of these vectors). This means that by Result 29 (VI.1 • 1), our solution is

\[
z(t) = C \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{4t} + D \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{5t} = \begin{bmatrix} 2Ce^{4t} + De^{5t} \\ Ce^{4t} + De^{5t} \end{bmatrix}
\]

for some constants \( C \) and \( D \).

---

VI.4.A • 2.  Example 76

Solve the differential equation

\[
z' = Az = \begin{bmatrix} i + 1 & 0 \\ -2 & 3 \end{bmatrix} z
\]

for the vector function \( z(t) \).

\[Solution::\]

Proceed as before to identify the eigenvalues. We again solve \( \det(A - \lambda \text{Id}) = 0 \) for \( \lambda \).

\[
\det(A - \lambda \text{Id}) = 0 \quad \iff \quad \det \left( \begin{bmatrix} i + 1 - \lambda & 0 \\ -2 & 3 - \lambda \end{bmatrix} \right) = 0 \\
\iff (i + 1 - \lambda)(3 - \lambda) = 0 \\
\iff \lambda = 3 \text{, or } \lambda = i + 1.
\]

So we have eigenvalues 3 and \( i + 1 \). Now we must find some associated eigenvectors. To do this, we consider \((A - \lambda \text{Id})v = 0\) for each eigenvalue \( \lambda \), and solve for \( v \) in each case. This is accomplished by Gaussian elimination.
\[ \lambda = i + 1 \quad \rightarrow \quad \begin{bmatrix} i + 1 - \lambda & 0 \\ -2 & 3 - \lambda \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow \begin{bmatrix} 0 & 0 \\ -2 & 2 - i \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow -2v_1 + (2 - i)v_2 = 0 \]

\[ \leftrightarrow v_1 = \frac{(2 - i)}{2} v_2. \]

\[ \lambda = 3 \quad \rightarrow \quad \begin{bmatrix} i + 1 - \lambda & 0 \\ -2 & 3 - \lambda \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow \begin{bmatrix} i - 2 & 0 \\ -2 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow v_1 = 0. \]

Hence we get eigenvectors \((2 - i, 2)\) with eigenvalue \(i + 1\) and \((0, 1)\) with eigenvalue \(3\) (as well as any non-zero scaling of these vectors). This means that by Result 29 (VI.1 • 1), our solution is

\[ z(t) = C \begin{bmatrix} 2 - i \\ 2 \end{bmatrix} e^{(i+1)t} + D \begin{bmatrix} 0 \\ 1 \end{bmatrix} e^{3t}, \]

for some constants \(C\) and \(D\).

### VI.4.A • 3. Example 77

Solve the differential equation

\[ z' = Az = \begin{bmatrix} 8 & 4 \\ -4 & 6 \end{bmatrix} z \]

for the vector function \(z(t)\).

**Solution**:

Proceed as before to identify the eigenvalues. We again solve \(\det(A - \lambda \text{Id}) = 0\) for \(\lambda\).

\[ \det(A - \lambda \text{Id}) = 0 \quad \leftrightarrow \quad \det \begin{bmatrix} 8 - \lambda & 4 \\ -4 & 6 - \lambda \end{bmatrix} = 0 \]

\[ \leftrightarrow (8 - \lambda)(6 - \lambda) + 16 = 0 \]

\[ \leftrightarrow \lambda^2 - 14\lambda + 64 \]

\[ \leftrightarrow \lambda = \frac{14 \pm \sqrt{14^2 - 4 \cdot 64}}{2} = 7 \pm i\sqrt{15}. \]

So we have eigenvalues \(7 + i\sqrt{15}\) and \(7 - i\sqrt{15}\). Now we must find some associated eigenvectors. To do this, we consider \((A - \lambda \text{Id})v = 0\) for each eigenvalue \(\lambda\), and solve for \(v\) in each case. This is accomplished by Gaussian elimination.

\[ \lambda = 7 + i\sqrt{15} \quad \rightarrow \quad \begin{bmatrix} 8 - \lambda & 4 \\ -4 & 6 - \lambda \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow \begin{bmatrix} 1 - i\sqrt{15} & 4 \\ -4 & -1 - i\sqrt{15} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow \begin{bmatrix} 16 & 4(1 + i\sqrt{15}) \\ -4 & -(1 + i\sqrt{15}) \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]

\[ \leftrightarrow \begin{bmatrix} 4 & 1 + i\sqrt{15} \\ -4 & -(1 + i\sqrt{15}) \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \]
We get eigenvectors

\[ \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \]

with eigenvalue \( 7 + i \sqrt{15} \); and

\[ \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \]

with eigenvalue \( 7 - i \sqrt{15} \)

as well as any non-0 scaling of these vectors. This means that by Result 30 (VI.1 • 2), our solution is

\[ z(t) = C e^{7t} \left( \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \right) \cos(\sqrt{15}t) - \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \sin(\sqrt{15}t) + D e^{7t} \left( \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \right) \sin(\sqrt{15}t) + \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \cos(\sqrt{15}t) \]

for some constants \( C \) and \( D \).

**VI.4.A • 4. Example 78 (Two Eigenvalues)**

Solve the system of differential equations

\[ \begin{aligned} x' &= 3x + 2y \\ y' &= 2x + 3y, \end{aligned} \]

with initial conditions \( x(0) = 2 \), and \( y(0) = 12 \).

**Solution:**

Firstly, let’s transform this into a matrix equation:

\[ \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = Az. \]

The first step is to find the eigenvalues of \( A \):

\[ \det(A - \lambda \text{Id}) = \det\left( \begin{bmatrix} 3 & 2 \\ 2 & 3 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) \]
Solve the system of differential equations with initial conditions given by
which means any \( k \text{ det} A \text{ eigenvalues of } A \text{ are pretty clearly just } A \text{ eigenvalues. In matrix form, this says }
\begin{align*}
\lambda = 1 & \quad \Rightarrow (A - \lambda \text{ Id}) v = 0 \quad \Rightarrow \begin{bmatrix} 2 - \lambda & 2 \\ 2 & 3 - \lambda \end{bmatrix} v = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
\lambda = 5 & \quad \Rightarrow (A - \lambda \text{ Id}) v = 0 \quad \Rightarrow \begin{bmatrix} 2 - \lambda & 2 \\ 2 & 3 - \lambda \end{bmatrix} v = \begin{bmatrix} 0 \\ 0 \end{bmatrix} 
\end{align*}
where \( k_1 \neq 0 \text{ denotes a constant multiple. In particular, we have eigenvectors when } k_1 = k_2 = 1. \text{ Using } 2 \times 2 \text{ Systems (VI.1 • 3), we get }
\begin{align*}
z(t) &= C \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^t + D \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{5t} = \begin{bmatrix} Ce^t + De^{5t} \\ -Ce^t + De^{5t} \end{bmatrix}.
\end{align*}
We need the initial conditions to find \( C \) and \( D \).
\begin{align*}
\begin{bmatrix} 2 \\ 12 \end{bmatrix} &= \begin{bmatrix} x(0) \\ y(0) \end{bmatrix} = z(0) = \begin{bmatrix} Ce^0 + De^{0 \cdot 5} \\ -Ce^0 + De^{0 \cdot 5} \end{bmatrix} = \begin{bmatrix} C + D \\ -C + D \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix}.
\end{align*}
Solve for \( C \) and \( D \) as follows: \( 2 = C + D \) implies \( D = 2 - C \). So \( D - C = 12 \) implies \( 2 - 2C = 12 \), i.e. \( C = -5 \). Then \( D = 2 + 5 = 7 \). Hence the final answer is
\begin{align*}
z(t) &= -5 \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^t + 7 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{5t} = \begin{bmatrix} -5e^t + 7e^{5t} \\ 5e^t + 7e^{5t} \end{bmatrix}.
\end{align*}

VI.4.A • 5. Example 79 (One Eigenvalue, Two Eigenvectors)

Solve the system of differential equations
\[
\begin{align*}
x' &= 2x \\
y' &= 2y,
\end{align*}
\]
with initial conditions \( x(0) = 2 \), and \( y(0) = 1 \).

Proof

In this case, there is no interaction between \( x \) and \( y \). Ordinarily, we could just solve it as two linear, first-order, differential equations. But to become more familiar with the techniques, let’s use some ideas from linear algebra. In matrix form, this says
\[
\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = Ax.
\]
The eigenvalues of \( A \) are pretty clearly just 2: \( \text{det}(A - \lambda \text{ Id}) = (2 - \lambda)^2 \), which is 0 iff \( \lambda = 2 \). The eigenvectors, however, then must satisfy
\[
\begin{bmatrix} 0 \\ 0 \end{bmatrix} = (A - \lambda \text{ Id}) v = \begin{bmatrix} 2 - \lambda & 0 \\ 0 & 2 - \lambda \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix},
\]
which means any \( v_1 \) and \( v_2 \) yield an eigenvector. In particular, \( \langle v_1, v_2 \rangle = (1, 0) \), and \( \langle v_1, v_2 \rangle = (0, 1) \). These
are linearly independent, and so by \(2 \times 2\) Systems (VI.1 • 3), the solution is
\[
z(t) = C \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^{2t} + D \begin{bmatrix} 0 \\ 1 \end{bmatrix} e^{2t} = \begin{bmatrix} Ce^{2t} \\ De^{2t} \end{bmatrix}.
\]

With the initial conditions \(x(0) = 2\) and \(y(0) = 1\), we quickly see that the solution to our initial value problem is \(\langle x(t), y(t) \rangle = \langle 2e^{2t}, e^{2t} \rangle\).

VI.4.A • 6. Example 80 (One Eigenvalue, One Eigenvector)

Solve the system of differential equations
\[
x' = x + 4y
\]
\[
y' = -4x + 9y,
\]
with initial conditions \(x(0) = -1\), and \(y(0) = 0\).

Solution ::

Written in matrix form, we get
\[
z' = \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & 4 \\ -4 & 9 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = Az.
\]

The eigenvalues will be \(\lambda\) where \(0 = (1 - \lambda)(9 - \lambda) + 16 = \lambda - 10\lambda + 25 = (5 - \lambda)^2\), i.e. \(\lambda = 5\). So we only have one eigenvalue. The critical question is what the eigenvectors then look like.
\[
\begin{bmatrix} 1 - \lambda & 4 \\ -4 & 9 - \lambda \end{bmatrix} v_1 = 0
\]
\[
\begin{bmatrix} -4 & 4 \\ -4 & 4 \end{bmatrix} v_1 = 0
\]
\[
\leftrightarrow v_1 = v_2 \leftrightarrow v = k \begin{bmatrix} 1 \\ 1 \end{bmatrix}
\]

for any \(k \neq 0\). Hence we only have one eigenvector up to scalars. Thus we must solve for \(m\) below: take the eigenvector \(v = \langle 1, 1 \rangle\).

\[
(A - \lambda I) m = v \leftrightarrow \begin{bmatrix} 1 - \lambda & 4 \\ -4 & 9 - \lambda \end{bmatrix} m_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \leftrightarrow \begin{bmatrix} -4 & 4 \\ -4 & 4 \end{bmatrix} m_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}
\]
\[
\leftrightarrow -4m_1 + 4m_2 = 1 \leftrightarrow m_2 = \frac{1}{4} + m_1
\]
\[
\leftrightarrow m = \begin{bmatrix} m_1 \\ m_1 + 1/4 \end{bmatrix} = \begin{bmatrix} m_1 \\ m_1 \end{bmatrix} + \begin{bmatrix} 0 \\ 1/4 \end{bmatrix}.
\]

Hence for \(m_1 = 0\), we get \(m = \langle 0, 1/4 \rangle\), and so by \(2 \times 2\) Systems (VI.1 • 3),
\[
z(t) = C \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{5t} + D \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{5t} + \begin{bmatrix} 0 \\ 1/4 \end{bmatrix} e^{5t}
\]
\[
= \begin{bmatrix} Ce^{5t} + Dte^{5t} \\ (C + D/4)e^{5t} + Dte^{5t} \end{bmatrix}.
\]

Using our initial conditions, we can solve for \(C\) and \(D\):
\[
\begin{bmatrix} -1 \\ 0 \end{bmatrix} = \begin{bmatrix} x(0) \\ y(0) \end{bmatrix} = z(0) = \begin{bmatrix} C \\ (C + D/4) \end{bmatrix}
\]
\[
\therefore \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} -1 \\ 4 \end{bmatrix}.
\]

Thus the final answer is
\[
z(t) = -1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{5t} + 4 \begin{bmatrix} 1 \\ 1 \end{bmatrix} t e^{5t} + \begin{bmatrix} 0 \\ 1/4 \end{bmatrix} e^{5t}
\]
VI.4.A • 7. Example 81 (Complex Eigenvalues)

Solve the initial value problem

\[ z' = \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 2 & 4 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = Az, \]

where \( x(0) = -4 \), and \( y(0) = 1 \).

**Solution:**

First we find the eigenvalues of \( A \):

\[
\det \left( \begin{bmatrix} 2 - \lambda & 4 \\ -1 & 2 - \lambda \end{bmatrix} \right) = (2 - \lambda)^2 + 4 = \lambda^2 - 4\lambda + 8
\]

\[
\lambda^2 - 4\lambda + 8 = 0 \iff \lambda = \frac{4 \pm \sqrt{16 - 4 \cdot 8}}{2} = 2 \pm 2i.
\]

Thus our eigenvalues are \( 2 + 2i \) and \( 2 - 2i \). We only need to deal with one eigenvalue in this case, since the matrix has only real numbers: the other eigenvalue will be the conjugate.

Let’s proceed with \( \lambda = 2 + 2i \) by Gaussian elimination to solve \((A - \lambda I)v = 0\):

\[
\begin{bmatrix} 2 - \lambda & 4 \\ -1 & 2 - \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \iff \begin{bmatrix} -2i & 4 \\ -1 & -2i \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ -1 & -2i \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix}
\]

So we get \(-v_1 - 2iv_2 = 0\), i.e. \( v = (v_1, v_2) = (-2i, 1)v_2 \). So, up to scalar multiplication, we get

\[
v_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} + i \begin{bmatrix} -2 \\ 0 \end{bmatrix}.
\]

Hence by Definition 1 (I.3 • 1), the other eigenvector is

\[
v_2 = \overline{v_1} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} + i \begin{bmatrix} 2 \\ 0 \end{bmatrix},
\]

with eigenvalue \( 2 - 2i \). Thus, by Definition 1 (I.3 • 1), the solution will be of the form

\[
z(t) = Ce^{2it}\begin{bmatrix} 0 \\ 1 \end{bmatrix} \cos(2t) - \begin{bmatrix} -2 \\ 0 \end{bmatrix} \sin(2t) + De^{2it}\begin{bmatrix} 0 \\ 1 \end{bmatrix} \sin(2t) + \begin{bmatrix} -2 \\ 0 \end{bmatrix} \cos(2t)
\]

\[
= \begin{bmatrix} 2Ce^{2it} \sin(2t) - 2De^{2it} \cos(2t) \\ Ce^{2it} \cos(2t) + De^{2it} \sin(2t) \end{bmatrix}.
\]

Using the initial conditions, we can solve for \( C \) and \( D \), and so arrive at our final answer.

\[
\begin{bmatrix} -4 \\ 1 \end{bmatrix} = \begin{bmatrix} C \cdot 0 - 2D \\ C + D \cdot 0 \end{bmatrix}
\]

\[
\therefore \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}
\]

\[
\therefore z(t) = \begin{bmatrix} 2e^{2t} \sin(2t) - 4e^{2t} \cos(2t) \\ e^{2t} \cos(2t) + 2e^{2t} \sin(2t) \end{bmatrix}.
\]
Solve the system of equations
\[
x' = 3x - y \\
y' = x + 2y,
\]
where \( x(0) = 0 \), and \( y(0) = 1 \).

**Solution:**

Representing this as a \( 2 \times 2 \) system with \( z = \langle x, y \rangle \), we get
\[
z' = \begin{bmatrix} 3 & -1 \\ 1 & 2 \end{bmatrix} z = Az,
\]
and so solve this we must first find the eigenvalues of \( A \) by solving \( \det(A - \lambda I) = 0 \) for \( \lambda \).
\[
\det(A - \lambda I) = (3 - \lambda)(2 - \lambda) + 1 = 7 - 5\lambda + \lambda^2.
\]
This is 0 iff, by the quadratic formula,
\[
\lambda = \frac{5 \pm i\sqrt{3}}{2}.
\]
Now we must find the eigenvectors of \( A \) by solving \( (A - \lambda I)v = 0 \) for \( v \), given each \( \lambda \) we found.
\[
\lambda = \frac{5 + i\sqrt{3}}{2} \rightarrow \begin{bmatrix} 3 - \lambda & -1 \\ 1 & 2 - \lambda \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0 \\
\rightarrow \begin{bmatrix} 1 - i\sqrt{3} \\ 2 \end{bmatrix} \Rightarrow \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0,
\]
After scaling the first row by \( \frac{2}{1-i\sqrt{3}} = \frac{1+i\sqrt{3}}{2} \), these will say the same thing: \( v_1 = (1/2 + i\sqrt{3}/2)v_2 \). Any choice of \( v_2 \neq 0 \) yields an eigenvector associated with this eigenvalue. So choose \( v_2 = 2 \), yielding an eigenvector
\[
v = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 1 + i\sqrt{3} \\ 2 \end{bmatrix} = \begin{bmatrix} 1/2 \\ i \sqrt{3} \\ 0 \end{bmatrix}.
\]
We could go through the same process for \( \lambda = (5 - i\sqrt{3})/2 \), but we’ll just end up with the conjugate of the above eigenvector. In either case, we can use previous results, either directly or by expansion with Euler’s formula, to get the general solution, writing \( \alpha \) for \( \sqrt{3}/2 \), of
\[
z(t) = \begin{bmatrix} z(t) \\ y(t) \end{bmatrix} = Ce^{5t/2} \begin{bmatrix} 1/2 \\ 0 \end{bmatrix} \cos(\alpha t) - \begin{bmatrix} \sqrt{3}/2 \\ \alpha \end{bmatrix} \sin(\alpha t) + De^{5t/2} \begin{bmatrix} 1/2 \\ 0 \end{bmatrix} \sin(\alpha t) + \begin{bmatrix} \sqrt{3}/2 \\ \alpha \end{bmatrix} \cos(\alpha t)
\]
\[
e^{5t/2} \begin{bmatrix} (C + \sqrt{3}D) \cos(\alpha t) + (D - \sqrt{3}C) \sin(\alpha t) \\ 2C \cos(\alpha t) + 2D \sin(\alpha t) \end{bmatrix}
\]
The initial conditions then tell us that
\[
\begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} C + \sqrt{3}D \\ 2C \end{bmatrix},
\]
so that \( C = 1/2 \), and \( D = -1/(2\sqrt{3}) = -\sqrt{3}/6 \). Hence the solution to the initial value problem is
\[
x(t) = \frac{2}{\sqrt{3}} e^{5t/2} \sin \left( \frac{\sqrt{3}}{2} t \right)
\]
\[
y(t) = e^{5t/2} \cos \left( \frac{\sqrt{3}}{2} t \right) - \frac{1}{\sqrt{3}} e^{5t/2} \sin \left( \frac{\sqrt{3}}{2} t \right).
\]
VI.4.A • 9. Example 83

Solve the system of equations

\[ \begin{align*}
    x' &= x - 2y \\
    y' &= 2x + y,
\end{align*} \]

with initial conditions \( x(0) = 1 \), and \( y(0) = 1 \).

**Solution:**

Firstly, transform this into the matrix equation

\[
\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}.
\]

Our first consideration is with the eigenvalues of \( A \).

\[
\begin{align*}
    \det(A - \lambda \text{Id}) &= (1 - \lambda)^2 + 4 = 5 - 2\lambda + \lambda^2, \\
    \text{which is } 0 \text{ iff } \lambda &= 1 \pm 2i.
\end{align*}
\]

Now we must find the eigenvalues. Firstly, consider \( \lambda = 1 + 2i \).

\[
\begin{align*}
    (A - \lambda \text{Id})v &= 0 \\
    \begin{bmatrix} -2i & -2 \\ 2 & -2i \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
    \begin{bmatrix} 1 & -i \\ 2 & -2i \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
    \begin{bmatrix} 1 & -i \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\
    v_1 &= iv_2 \\
    v &= \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 1 \\ i \end{bmatrix} v_2 = \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} + i \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) v_2.
\end{align*}
\]

So for any \( v_2 \neq 0 \), we get an eigenvector. So choose \( v_2 = 1 \) for simplicity. This yields the eigenvector \( \begin{bmatrix} 1 \\ i \end{bmatrix} \).

The other eigenvalue \( 1 - 2i \) is the conjugate of the first and thus has an eigenvector which is the conjugate of the first: \( \begin{bmatrix} 1 \\ -i \end{bmatrix} \).

Thus we can write that the general solution to the differential equation is

\[
\begin{align*}
    \begin{bmatrix} x(t) \\ y(t) \end{bmatrix} &= Ce^{t} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \cos(2t) - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \sin(2t) + De^{t} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \sin(2t) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \cos(2t) \\
    &= Ce^{t} \cos(2t) + De^{t} \sin(2t) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + De^{t} \cos(2t) \begin{bmatrix} 0 \\ 1 \end{bmatrix}.
\end{align*}
\]

The initial conditions immediately tell us \( C = D = 1 \), and thus gives the final answer:

\[
\begin{align*}
    \begin{bmatrix} x(t) \\ y(t) \end{bmatrix} &= e^{t} \begin{bmatrix} \cos(2t) + e^{t} \sin(2t) \\ e^{t} \cos(2t) - e^{t} \sin(2t) \end{bmatrix}.
\end{align*}
\]

VI.4.A • 10. Example 84

Solve the non-homogeneous system of differential equations

\[ \begin{align*}
    x' &= x - y + \sin(t) \\
    y' &= y + e^{2t},
\end{align*} \]

**Solution:**

Write this in matrix form:

\[
\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \sin(t) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} e^{2t}.
\]
First solving the homogeneous version

\[
\begin{bmatrix}
  x' \\
  y'
\end{bmatrix}
= \begin{bmatrix}
  1 & -1 \\
  0 & 1
\end{bmatrix}
\begin{bmatrix}
  x \\
  y
\end{bmatrix}
= Az.
\]

The eigenvalues of A are given by

\[
\det(A - \lambda I) = (1 - \lambda)^2,
\]

in other words, just \( \lambda = 1 \). The eigenvectors of this are just multiples of \( v = \langle 1, 0 \rangle \). Now we must find the generalized eigenvector \( m \) satisfying

\[
(A - \lambda I) m = \begin{bmatrix} 1 \\ 0 \end{bmatrix},
\]

\[
\begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix}
\begin{bmatrix} m_1 \\ m_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}
\]

\[
\rightarrow m_2 = -1.
\]

so \( m_1 \) can be anything as long as \( m_2 = -1 \). So we can choose, for example, \( m = \langle 0, -1 \rangle \). In this case, the general homogeneous solution is

\[
C \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^t + D \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} t e^t + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^t \right).
\]

Now we must find a particular solution to the non-homogeneous equation. To do this, the non-homogeneous part is a mix of sines, and exponentials. As a result, our guess shall be

\[
Z(t) = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \sin(t) + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \cos(t) + \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} e^{2t}
\]

\[
Z'(t) = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \cos(t) + \begin{bmatrix} -b_1 \\ -b_2 \end{bmatrix} \sin(t) + \begin{bmatrix} 2c_1 \\ 2c_2 \end{bmatrix} e^{2t}
\]

\[
A \cdot Z(t) + G(t) = \begin{bmatrix} a_1 - a_2 + 1 \\ a_2 \end{bmatrix} \sin(t) + \begin{bmatrix} b_1 - b_2 \\ b_2 \end{bmatrix} \cos(t) + \begin{bmatrix} c_1 - c_2 \\ c_2 + 1 \end{bmatrix} e^{2t}.
\]

Thus to satisfy the equation, we require \( Z'(t) = A \cdot Z(t) + G(t) \). By equating coefficients, this means that

\[
\begin{bmatrix} a_1 - a_2 + 1 \\ a_2 \end{bmatrix} = \begin{bmatrix} -b_1 \\ -b_2 \end{bmatrix},
\]

\[
\begin{bmatrix} b_1 - b_2 \\ b_2 \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix},
\]

\[
\begin{bmatrix} 2c_1 \\ 2c_2 \end{bmatrix} = \begin{bmatrix} c_1 - c_2 \\ c_2 + 1 \end{bmatrix}.
\]

The bottom row tells us that \(-b_2 = b_2\), and so \( a_2 = b_2 = 0 \). The bottom row also tells us that \( 2c_2 = c_2 + 1 \), i.e. \( c_2 = 1 \). This leaves us with

\[
\begin{bmatrix} a_1 + 1 \\ a_2 \end{bmatrix} = \begin{bmatrix} -b_1 \\ -b_2 \end{bmatrix},
\]

\[
\begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix},
\]

\[
\begin{bmatrix} 2c_1 \\ 2c_2 \end{bmatrix} = \begin{bmatrix} c_1 - 1 \\ 2 \end{bmatrix}.
\]

The top row tells us that \( a_1 + 1 = -a_1 \), in other words, \( a_1 = -1/2 \), and \( b_1 = 1/2 \). The top row also tells us that \( c_1 = -1 \). Thus we have determined all the constants:

\[
Z(t) = \begin{bmatrix} -1/2 \\ 0 \end{bmatrix} \sin(t) + \begin{bmatrix} 1/2 \\ 0 \end{bmatrix} \cos(t) + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{2t}.
\]

Hence the general solution to the non-homogeneous differential equation is

\[
z(t) = C \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^t + D \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} t e^t + \begin{bmatrix} 0 \\ -1 \end{bmatrix} e^t \right) + \begin{bmatrix} -1/2 \\ 0 \end{bmatrix} \sin(t) + \begin{bmatrix} 1/2 \\ 0 \end{bmatrix} \cos(t) + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{2t}
\]

\[
= \begin{bmatrix} Ce^t + Dte^t - \sin(t)/2 + \cos(t)/2 - e^{2t} \\ -De^t + e^{2t} \end{bmatrix}
\]

for constants \( C \) and \( D \).
Find the general solution to the system
\[
\begin{bmatrix}
x' \\
y'
\end{bmatrix} = \begin{bmatrix} x + y + t \\ x + 2y \end{bmatrix}.
\]

**Solution:**

Represent this as a linear system:
\[
z' = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} z + \begin{bmatrix} 1 \\ 0 \end{bmatrix} t.
\]

First we solve the homogeneous part:
\[
z' = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} z.
\]

To do this, find the eigenvalues, and then the eigenvectors. Skipping these steps we get eigenvalues of \( \lambda_1 = \left(3 + \sqrt{5}\right)/2 \) and \( \lambda_2 = \left(3 - \sqrt{5}\right)/2 \), and eigenvectors of
\[
v_1 = \begin{bmatrix} 2 + \sqrt{5} \\ 1 \end{bmatrix}, \quad v_2 = \begin{bmatrix} 2 - \sqrt{5} \\ 1 \end{bmatrix}.
\]

Hence the general solution to the homogeneous equation is
\[
\Phi(t) = C \begin{bmatrix} 2 + \sqrt{5} \\ 1 \end{bmatrix} e^{\left(3 + \sqrt{5}\right)t/2} + D \begin{bmatrix} 2 - \sqrt{5} \\ 1 \end{bmatrix} e^{\left(3 - \sqrt{5}\right)t/2}.
\]

To find the non-homogeneous part, proceed by undetermined coefficients. Given the form \( \vec{c} t \), our guess will be \( \vec{Z} t = \vec{a} t + \vec{b} \). We thus need the following to be equal:

\[
\begin{align*}
AZ + G(t) &= \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} + \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} t \\
&= \begin{bmatrix} a_1 + a_2 + 1 \\ a_1 + 2a_2 \end{bmatrix} t + \begin{bmatrix} b_1 + b_2 \\ b_1 + 2b_2 \end{bmatrix}.
\end{align*}
\]

Thus by equating coefficients, we must solve
\[
\begin{align*}
[a_1 \\ a_2] &= \begin{bmatrix} b_1 + b_2 \\ b_1 + 2b_2 \end{bmatrix}, \quad \text{and} \quad [a_1 + a_2 + 1 \\ a_1 + 2a_2] = \begin{bmatrix} 0 \\ 0 \end{bmatrix}
\end{align*}
\]

The equations on the right tell us that \( a_1 = -2a_2 \), and thus by the first, \( a_2 = 1 \) so that \( a_1 = -2 \). Using this, the equations on the left tell us that
\[
\begin{bmatrix} -2 \\ 1 \end{bmatrix} = \begin{bmatrix} b_1 + b_2 \\ b_1 + 2b_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}.
\]

Applying the inverse of the matrix to both sides (or just solving the two equations for \( b_1 \) and \( b_2 \)) yields that \( b_1 = -5 \) and \( b_2 = 3 \). Hence the general solution to the non-homogeneous equation is
\[
z = \Phi(t) + Z(t)
\]
\[
= C \begin{bmatrix} 2 + \sqrt{5} \\ 1 \end{bmatrix} e^{\left(3 + \sqrt{5}\right)t/2} + D \begin{bmatrix} 2 - \sqrt{5} \\ 1 \end{bmatrix} e^{\left(3 - \sqrt{5}\right)t/2} + \begin{bmatrix} -2 \\ 1 \end{bmatrix} t + \begin{bmatrix} -5 \\ 3 \end{bmatrix}
\]
\[
= \begin{bmatrix} 2Ce^{\left(3 + \sqrt{5}\right)t/2} + \frac{2De^{\left(3 - \sqrt{5}\right)t/2}}{1 - \sqrt{5}} - 2t - 5 \\ Ce^{\left(3 + \sqrt{5}\right)t/2} + De^{\left(3 - \sqrt{5}\right)t/2} + t + 3 \end{bmatrix}.
\]
### Example 86

Consider the non-linear system of differential equations

\[
\begin{bmatrix}
    x' \\
    y'
\end{bmatrix} = \begin{bmatrix}
    x + \cos(y) \\
    x \cdot y
\end{bmatrix} = \begin{bmatrix}
    f(x, y) \\
    g(x, y)
\end{bmatrix}.
\]

1. Find the critical points of the system;
2. give the linearization of this around the point \( s = (0, \pi/2) \);
3. classify the stability that \( s \) has.

**Solution:**

1. The critical points of the system are when both \( x + \cos(y) \) and \( x \cdot y \) are 0. From \( x \cdot y = 0 \), we get that \( x = 0 \) or \( y = 0 \). If \( y = 0 \), then \( x + \cos(y) = x + 1 \), which is 0 if \( x = -1 \). Otherwise \( x = 0 \), and so \( \cos(y) = 0 \), requiring \( y = k\pi/2 \) for \( k \) an odd integer.

So the critical points are \( (1, 0) \) and \( (0, 2n + 1)\pi/2 \) for any integer \( n \).

2. To find the linearization around any given point, first we must calculate the partial derivatives of \( f(x, y) \) and \( g(x, y) \).

\[
\begin{align*}
    f_x(x, y) &= 1, & f_y(x, y) &= -\sin y, \\
    g_x(x, y) &= y, & g_y(x, y) &= x.
\end{align*}
\]

This yields the jacobian matrix evaluated at \( s = (0, \pi/2) \) as

\[
J(F)(s) = \begin{bmatrix}
    f_x(s) & f_y(s) \\
    g_x(s) & g_y(s)
\end{bmatrix} = \begin{bmatrix}
    1 & -\sin(\pi/2) \\
    \pi/2 & 0
\end{bmatrix} = \begin{bmatrix}
    1 & -1 \\
    \pi/2 & 0
\end{bmatrix}.
\]

Hence the linearization is, with \( u = x \) and \( v = y - \pi/2 \),

\[
\begin{bmatrix}
    u' \\
    v'
\end{bmatrix} = \begin{bmatrix}
    1 & -1 \\
    \pi/2 & 0
\end{bmatrix} \begin{bmatrix}
    u \\
    v
\end{bmatrix}.
\]

In principle, we know how to find the general solution to this, but actually doing this is messy of course, since it will involve roots of \( \pi \) and so forth.

3. Use Result 36 (VI.3 • 3) to pass the question onto the stability of the linearization found in (2). To do this, we must find the eigenvalues of \( J(F)(s) \).

\[
0 = \det(J(F)(s) - \lambda \text{Id}) \iff 0 = \det\left(\begin{bmatrix}
    1 - \lambda & -1 \\
    \pi/2 & -\lambda
\end{bmatrix}\right) = -\lambda(1 - \lambda) + \pi/2
\]

\[
\iff 0 = \lambda^2 - \lambda + \pi/2
\]

\[
\iff \lambda = \frac{1 \pm \sqrt{1 - 2\pi}}{2} = \frac{1 \pm i\sqrt{2\pi - 1}}{2}.
\]

so that the real part is positive for each eigenvalue. Hence the critical point \( w = 0 \) is unstable, and so the critical point \( z = s \) is unstable.

### Example 87

Consider the following non-linear system

\[
\begin{align*}
    x' &= 3x - y + 3y^2, \\
    y' &= x + 2y
\end{align*}
\]

with initial conditions \( x(0) = 1 \), and \( y(0) = 1 \).

1. Find the critical points of the system;
2. Find the linear approximation around each point.
3. One critical point is \( (0, 0) \). What is the type of this critical point?

**Note:** you only need to find the eigenvalues to answer this part.
Solution .:

1. Critical points \( (a, b) \) will satisfy

\[
0 = 3a - b + 3b^2,
0 = a + 2b = 0.
\]

The second requires that \( a = -2b \). Using this in the first equation tells us that \( 0 = -7b + 3b^2 = (-7 + 3b)b \) so that \( b = 0 \) or \( b = 7/3 \). Hence we have the critical points \( (0, 0) \), and \( (-14/3, 7/3) \).

2. First we must find the Jacobian of \( \langle f, g \rangle \) where \( f(x, y) = 3x - y + 3y^2 \) and \( g(x, y) = x + 2y \):

\[
J(F)(a, b) = \begin{bmatrix}
    f_x(a, b) & f_y(a, b) \\
    g_x(a, b) & g_y(a, b)
\end{bmatrix} = \begin{bmatrix}
    3 & -1 + 6b \\
    1 & 2
\end{bmatrix}.
\]

Hence the linearization around \( (-14/3, 7/3) \) is

\[
\begin{bmatrix}
    u' \\
    v'
\end{bmatrix} = \begin{bmatrix}
    3 & 13 \\
    1 & 2
\end{bmatrix} \begin{bmatrix}
    u \\
    v
\end{bmatrix},
\]

and around \( (0, 0) \), the linearization is

\[
\begin{bmatrix}
    u' \\
    v'
\end{bmatrix} = \begin{bmatrix}
    3 & -1 \\
    1 & 2
\end{bmatrix} \begin{bmatrix}
    u \\
    v
\end{bmatrix}.
\]

3. First we must find the eigenvalues:

\[
\det \left( \begin{bmatrix}
    3 - \lambda & -1 \\
    1 & 2 - \lambda
\end{bmatrix} \right) = (3 - \lambda)(2 - \lambda) + 1 = \lambda^2 - 5\lambda + 7,
\]

which is 0 iff \( \lambda = \frac{5 \pm \sqrt{3}}{2} \). The real part of these is positive, and so this critical point is unstable.

§ VI.4.B. \( 3 \times 3 \) Systems

Now let’s consider some \( 3 \times 3 \) systems, which are computationally more involved. As a result, many of the algebraic steps will be skipped.

### VI.4.B • 1. Example 88 (Three Eigenvalues)

Solve the initial value problem

\[
\begin{bmatrix}
    x' \\
    y' \\
    z'
\end{bmatrix} = \begin{bmatrix}
    1 & 14 & 12 \\
    -1 & 10 & 6 \\
    1 & -11 & -7
\end{bmatrix} \begin{bmatrix}
    x \\
    y \\
    z
\end{bmatrix} = Az,
\]

where \( z(0) = (0, 3, 1) \).

Solution .:

The first step is to find the eigenvalues of \( A \) by solving \( \det(A - \lambda \text{Id}) = 0 \) for \( \lambda \).

\[
\det \left( \begin{bmatrix}
    1 - \lambda & 14 & 12 \\
    -1 & 10 - \lambda & 6 \\
    1 & -11 & -7 - \lambda
\end{bmatrix} \right)
\]

\[
= (1 - \lambda)(10 - \lambda)(-7 - \lambda) - 1 \cdot 6 \cdot (-11) + 14 \cdot 6 \cdot 1
- 14 \cdot (-1) \cdot (-7 - \lambda) + 12 \cdot (-1) \cdot (-11) - 12 \cdot (10 - \lambda) \cdot 1
= -\lambda^3 + 4\lambda^2 - \lambda - 6 = (\lambda - 3)(\lambda - 2)(\lambda + 1).
\]

This is 0 iff \( \lambda = 3, \lambda = 2, \) or \( \lambda = -1 \). So 3, 2, and -1 are the eigenvalues of \( A \). Now we want to find the eigenvectors. We do this by solving the equation \((A - \lambda \text{Id})v = 0\) for each of our eigenvalues \( \lambda \).
First let's deal with $\lambda = 3$. We start with the augmented matrix, and proceed by Gaussian elimination:

$$
\begin{bmatrix}
1 - \lambda & 14 & 12 & 0 \\
-1 & 10 - \lambda & 6 & 0 \\
1 & -11 & -7 - \lambda & 0 \\
\end{bmatrix}
\begin{bmatrix}
-2 & 14 & 12 & 0 \\
-1 & 7 & 6 & 0 \\
1 & -11 & -10 & 0 \\
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & -11 & -10 & 0 \\
0 & -4 & -4 & 0 \\
0 & 0 & 0 & 0 \\
\end{bmatrix}
\begin{bmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 1 & 0 \\
0 & 0 & 0 & 0 \\
\end{bmatrix}
$$

Hence we get the eigenvector $v = \langle v_1, v_2, v_3 \rangle$ with $v_1 + v_3 = 0$, and $v_2 + v_3 = 0$. In other words, we get the eigenvector $v = \langle 1, 1, -1 \rangle$ up to scaling.

Now we deal with $\lambda = 2$. We again start with the augmented matrix, and proceed by Gaussian elimination:

$$
\begin{bmatrix}
1 - \lambda & 14 & 12 & 0 \\
-1 & 10 - \lambda & 6 & 0 \\
1 & -11 & -7 - \lambda & 0 \\
\end{bmatrix}
\begin{bmatrix}
2 & 14 & 12 & 0 \\
-1 & 11 & 6 & 0 \\
1 & -11 & -6 & 0 \\
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & -14 & -12 & 0 \\
0 & -6 & -6 & 0 \\
0 & 3 & 3 & 0 \\
\end{bmatrix}
\begin{bmatrix}
1 & 0 & 2 & 0 \\
0 & 1 & 1 & 0 \\
0 & 0 & 0 & 0 \\
\end{bmatrix}
$$

Hence we get the eigenvector $v = \langle v_1, v_2, v_3 \rangle$ with $v_1 + 2v_3 = 0$, and $v_2 + v_3 = 0$. In other words, we get the eigenvector $v = \langle 2, 1, -1 \rangle$ up to scaling.

Now we deal with the last eigenvalue $\lambda = -1$. Again, proceed by Gaussian elimination:

$$
\begin{bmatrix}
1 - \lambda & 14 & 12 & 0 \\
-1 & 10 - \lambda & 6 & 0 \\
1 & -11 & -7 - \lambda & 0 \\
\end{bmatrix}
\begin{bmatrix}
2 & 14 & 12 & 0 \\
-1 & 11 & 6 & 0 \\
1 & -11 & -6 & 0 \\
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 7 & 6 & 0 \\
0 & 18 & 12 & 0 \\
0 & 0 & 0 & 0 \\
\end{bmatrix}
\begin{bmatrix}
1 & 0 & 4/3 & 0 \\
0 & 1 & 2/3 & 0 \\
0 & 0 & 0 & 0 \\
\end{bmatrix}
$$

Hence we get the eigenvector $v = \langle v_1, v_2, v_3 \rangle$ with $v_1 + 4/3v_3 = 0$, and $v_2 + 2/3v_3 = 0$. In other words, we get the eigenvector $v = \langle 4, 2, -3 \rangle$. Hence by (VI.1 • 5), the solution is of the form

$$z(t) = C_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{3t} + C_2 \begin{bmatrix} 2 \\ -1 \end{bmatrix} e^{2t} + C_3 \begin{bmatrix} 4 \\ -3 \end{bmatrix} e^{-t}.$$ 

With the initial condition of $z(t) = \langle 0, 3, 1 \rangle$, we can solve for $C_1, C_2, C_3$:

$$
\begin{bmatrix}
0 \\
3 \\
1 \\
\end{bmatrix}
= C_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + C_2 \begin{bmatrix} 2 \\ -1 \end{bmatrix} + C_3 \begin{bmatrix} 4 \\ -3 \end{bmatrix}
= \begin{bmatrix} 1 & 2 & 4 \\ 1 & 1 & 2 \\ -1 & -1 & -3 \end{bmatrix} \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix}

\therefore \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix} = \begin{bmatrix} 6 \\ 5 \\ -4 \end{bmatrix}
$$

Thus we can write the solution to the initial value problem as

$$z(t) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{3t} + 6 \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{3t} + 5 \begin{bmatrix} 2 \\ -1 \end{bmatrix} e^{2t} - 4 \begin{bmatrix} 2 \\ -3 \end{bmatrix} e^{-t} = \begin{bmatrix} 6 e^{3t} + 10 e^{2t} - 16 e^{-t} \\ 6 e^{3t} + 5 e^{2t} - 8 e^{-t} \\ -6 e^{-t} - 5 e^{3t} + 12 e^{-t} \end{bmatrix}.$$
Solution:

The first step here is to find the eigenvalues.

\[
\det(A - \lambda \text{Id}) = \det \begin{bmatrix}
1 - \lambda & 0 & 2 \\
-1 & -1 - \lambda & 2 \\
0 & 16 & -1 - \lambda 
\end{bmatrix}
\]

\[
= -\lambda^3 - \lambda^2 + 33\lambda - 63
\]

\[
= -(\lambda + 7)(\lambda - 3)^2.
\]

Hence the eigenvalues are \(-7\) and \(3\). Now we must find the eigenvectors by solving \((A - \lambda \text{Id})v = \vec{0}\) for each eigenvalue \(\lambda\) that we found. We do this by Gaussian elimination.

\(\lambda = -7 \quad \Rightarrow \quad (A - \lambda \text{Id})v = \vec{0}\)

\[
\begin{bmatrix}
1 - \lambda & 0 & 2 \\
-1 & -1 - \lambda & 2 \\
0 & 16 & -1 - \lambda 
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
v_3 
\end{bmatrix} = 
\begin{bmatrix}
0 \\
0 \\
0 
\end{bmatrix}
\]

\[
\Rightarrow
\begin{bmatrix}
4 & 0 & 1 \\
0 & 8 & 3 \\
-4 & 24 & 8 
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
v_3 
\end{bmatrix} = 
\begin{bmatrix}
0 \\
0 \\
0 
\end{bmatrix}
\]

\[
\Rightarrow
\begin{bmatrix}
1 & 0 & 1/4 \\
0 & 1 & 3/8 \\
0 & 0 & 0 
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
v_3 
\end{bmatrix} = 
\begin{bmatrix}
0 \\
0 \\
0 
\end{bmatrix}
\]

\[
\Rightarrow v = v_3 \begin{bmatrix}
-1/4 \\
-3/8 \\
1 
\end{bmatrix}, \text{ e.g. for } v_3 = 8, \begin{bmatrix}
-2 \\
-3 \\
8 
\end{bmatrix}.
\]

\(\lambda = 3 \quad \Rightarrow \quad (A - \lambda \text{Id})v = \vec{0}\)

\[
\begin{bmatrix}
1 - \lambda & 0 & 2 \\
-1 & -1 - \lambda & 2 \\
0 & 16 & -1 - \lambda 
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
v_3 
\end{bmatrix} = 
\begin{bmatrix}
0 \\
0 \\
0 
\end{bmatrix}
\]

\[
\Rightarrow
\begin{bmatrix}
1 & 0 & -1 \\
0 & 1 & -1/4 \\
0 & 0 & 0 
\end{bmatrix}
\begin{bmatrix}
v_1 \\
v_2 \\
v_3 
\end{bmatrix} = 
\begin{bmatrix}
0 \\
v_3 \\
v_3 
\end{bmatrix}, \text{ e.g. for } v_3 = 4, \begin{bmatrix}
4 \\
1 \\
4 
\end{bmatrix}.
\]

Thus we don’t have linearly independent eigenvectors for the repeated eigenvalue \(\lambda_2 = \lambda_3 = 3\). Thus we must solve the equation \((A - \lambda_2 \text{Id})m = \vec{v}_2\), where \(\lambda_2 = 3\), and \(\vec{v}_2 = \langle 4, 1, 4 \rangle\) found above:

\[
\begin{bmatrix}
-2 & 0 & 2 \\
-1 & -4 & 2 \\
0 & 16 & -4 
\end{bmatrix}
\begin{bmatrix}
m_1 \\
m_2 \\
m_3 
\end{bmatrix} = 
\begin{bmatrix}
4 \\
1 \\
4 
\end{bmatrix}.
\]

Proceed by Gaussian elimination.

\[
\begin{bmatrix}
-2 & 0 & 2 & 4 \\
-1 & -4 & 2 & 1 \\
0 & 16 & -4 & 4 
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & -1 & -2 \\
-1 & -4 & 2 & 1 \\
0 & 4 & -1 & 1 
\end{bmatrix}
\Rightarrow
\begin{bmatrix}
1 & 0 & -1 & -2 \\
0 & -4 & 1 & -1 \\
0 & 4 & -1 & 1 
\end{bmatrix}
\]

\[\]

\[\]

\[\]

\[\]

\[\]
This then says that we can get an \( m_1 \) and \( m_2 \) that work given any \( m_3 \). To make things simpler, choose \( m_3 = 0 \).

In this case,
\[
\begin{bmatrix}
m_1 \\
m_2 \\
m_3
\end{bmatrix}
= \begin{bmatrix}
-2 \\
1/4 \\
0
\end{bmatrix},
\]
and so (VI.1 • 5)—in combination with its comment afterward—tells us that the solution is of the form
\[
z(t) = C_1 \begin{bmatrix}
-2 \\
3 \\
8
\end{bmatrix} e^{-7t} + C_2 \begin{bmatrix}
4 \\
1 \\
4
\end{bmatrix} e^{3t} + C_3 \begin{bmatrix}
4 \\
1 \\
0
\end{bmatrix} t e^{3t} + \begin{bmatrix}
-2 \\
1/4 \\
0
\end{bmatrix} e^{3t}.
\]

Using the initial condition that \( z(0) = \langle 200, 0, 100 \rangle \) gives us a matrix equation that we again solve for by Gaussian elimination.
\[
\begin{bmatrix}
200 \\
0 \\
100
\end{bmatrix}
= C_1 \begin{bmatrix}
-2 \\
3 \\
8
\end{bmatrix} + C_2 \begin{bmatrix}
4 \\
1 \\
4
\end{bmatrix} + C_3 \begin{bmatrix}
4 \\
1 \\
0
\end{bmatrix}
\]
\[
= \begin{bmatrix}
-2 & 4 & -2 \\
3 & 1 & 1/4 \\
8 & 4 & 0
\end{bmatrix}
\begin{bmatrix}
C_1 \\
C_2 \\
C_3
\end{bmatrix},
\]
\[
\begin{bmatrix}
C_1 \\
C_2 \\
C_3
\end{bmatrix}
= \begin{bmatrix}
2 \\
1/2 \\
-60
\end{bmatrix}.
\]

Thus we can write that the final answer is
\[
z(t) = 2 \begin{bmatrix}
-2 \\
3 \\
8
\end{bmatrix} e^{-7t} + 21 \begin{bmatrix}
4 \\
1 \\
4
\end{bmatrix} e^{3t} - 60 \begin{bmatrix}
4 \\
1 \\
0
\end{bmatrix} t e^{3t} + \begin{bmatrix}
-2 \\
1/4 \\
0
\end{bmatrix} e^{3t}
\]
\[
= -4e^{-7t} + 204e^{3t} - 240te^{3t}
\]
\[
= -6e^{-7t} + 6e^{3t} - 60te^{3t}
\]
\[
= 16e^{-7t} + 84e^{3t} - 240te^{3t}.
\]
Appendix A. Unnecessary Proofs

A.1. Result A1

Consider the system of linear differential equations $z' = A(t) \cdot z + G(t)$ written as

$$
\begin{bmatrix}
x' \\
y'
\end{bmatrix} =
\begin{bmatrix}
a(t) & b(t) \\
c(t) & d(t)
\end{bmatrix}
\begin{bmatrix}
x \\
y
\end{bmatrix} +
\begin{bmatrix}
g_1(t) \\
g_2(t)
\end{bmatrix}.
$$

If $b(t) \neq 0$ and $c(t) \neq 0$, we can write

$$
x'' - \left(a + d + \frac{b'}{b}\right)x' + \left(ad - bc - a' - \frac{b'}{b}\right)x = g_1' - \left(\frac{b'}{b} + d\right)g_1 + bg_2
$$

$$
y'' - \left(d + a + \frac{c'}{c}\right)y' + \left(da - cb - d' - \frac{c'}{c}\right)y = g_2' - \left(\frac{c'}{c} + a\right)g_2 + cg_1.
$$

In the case that $a, b, c, d$ are all constants, we have

$$
x'' - \text{trace}(A)x' + \det(A)x = g_1' - dg_1 + bg_2
$$

$$
y'' - \text{trace}(A)y' + \det(A)y = g_2' - ag_2 + cg_1.
$$

**Proof:**

Consider the system of linear differential equations $z = A(t)z + G(t)$, also written

$$
\begin{bmatrix}
x' \\
y'
\end{bmatrix} =
\begin{bmatrix}
a(t) & b(t) \\
c(t) & d(t)
\end{bmatrix}
\begin{bmatrix}
x \\
y
\end{bmatrix} +
\begin{bmatrix}
g_1(t) \\
g_2(t)
\end{bmatrix}.
$$

Understanding that $a, b, c, d$ are all functions of $t$, I will now abandon writing “$(t)$” for the sake of space. Now the above equality means, so long as we can divide by $b$ and $c$,

$$
y = \frac{x' - ax - g_1}{b}, \quad x = \frac{y' - dy - g_2}{c}.
$$

differentiation on the matrix equation yields

$$
z'' = A'z + Az' + G',
$$

which can be written

$$
\begin{bmatrix}
x'' \\
y''
\end{bmatrix} =
\begin{bmatrix}
a' & b' \\
c' & d'
\end{bmatrix}
\begin{bmatrix}
x' \\
y'
\end{bmatrix} +
\begin{bmatrix}
a & b \\
c & d
\end{bmatrix}
\begin{bmatrix}
x \\
y
\end{bmatrix} +
\begin{bmatrix}
g_1' \\
g_2'
\end{bmatrix}.
$$

Substituting the equalities of $y'$ in terms of $y$, then $y$ in terms of $x$ and $x'$ above, this gives the equations

$$
x'' = a'x + ax' + g_1 + b'y + by'
$$

$$
= a'x + ax' + g_1 + b'y + b(cx + dy + g_2)
$$

$$
= \left(a' + \frac{b'}{b} + cb - ad\right)x + \left(\frac{b'}{b} + a + d\right)x' + g_1' - \left(\frac{b'}{b} + d\right)g_1 + bg_2,
$$

and we get a similar result for $y$:

$$
y'' = \left(d' + \frac{c'}{c} + bc - da\right)y + \left(\frac{c'}{c} + d + a\right)y' + g_2' - \left(\frac{c'}{c} + a\right)g_2 + cg_1.
$$

In the case that $a, b, c, d$ are all constants, $a' = b' = c' = d' = 0$ and so this simplifies to the equations

$$
x'' - \text{trace}(A)x' + \det(A)x = g_1' - dg_1 + bg_2
$$

$$
y'' - \text{trace}(A)y' + \det(A)y = g_2' - ag_2 + cg_1.
$$
A • 2. Result A2

Let \( A \) be a \( 2 \times 2 \) matrix, and \( z(t) \) a vector. Consider the differential equation
\[
z' = A \cdot z.
\]
Suppose \( A \) has distinct eigenvalues \( \lambda_1 \) and \( \lambda_2 \) with some associated eigenvectors \( v \) and \( w \) respectively. Thus solutions to the above differential equation are of the form
\[
z(t) = C e^{\lambda_1 t} + D e^{\lambda_2 t},
\]
for some constants \( C \) and \( D \).

Proof:.

First of all, anything of that form \( C e^{\lambda_1 t} + D e^{\lambda_2 t} \) will be a solution to the differential equation. This is because
\[
z(t) = C e^{\lambda_1 t} + D e^{\lambda_2 t} \quad \longrightarrow \quad z'(t) = C \lambda_1 e^{\lambda_1 t} + D \lambda_2 e^{\lambda_2 t}
\]
\[
= C A \cdot v e^{\lambda_1 t} + D A \cdot w e^{\lambda_2 t}
\]
\[
= A \cdot (C e^{\lambda_1 t} + D e^{\lambda_2 t})
\]
\[
= A \cdot z(t).
\]
So now we need to show that any solution to \( z' = Az \) is of this form. Write out the matrix \( A \) and vector \( v \) as
\[
A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad v = \begin{bmatrix} x \\ y \end{bmatrix}.
\]
Thus we have
\[
x' = ax + by, \quad \text{and} \quad y' = cx + dy.
\]
Now we consider two cases:

(1) \( b = c = 0 \). In this case, the differential equations have no interaction: \( x' = ax \) and \( y' = dy \) so that the solution is
\[
z(t) = \begin{bmatrix} Ce^{at} \\ De^{dt} \end{bmatrix},
\]
for some constants \( C \) and \( D \). Note that this agrees with the predicted form of the result. To see this, the eigenvalues of \( A \) are \( a \) and \( d \). The eigen vectors are just \( \begin{bmatrix} 1 \\ 0 \end{bmatrix} \) times any constant, and \( \begin{bmatrix} 0 \\ 1 \end{bmatrix} \) times any constant. Hence the predicted solution is
\[
C \begin{bmatrix} c_1 \\ 0 \end{bmatrix} e^{at} + D \begin{bmatrix} 0 \\ c_2 \end{bmatrix} e^{dt} = \begin{bmatrix} C' e^{at} \\ D' e^{dt} \end{bmatrix}
\]
for some constants \( C, C', D, D', c_1, \) and \( c_2 \). And this is precisely the form it must be in above.

(2) \( b \neq 0 \) or \( c \neq 0 \). For definiteness, say \( b \neq 0 \), since it’s the same argument. This allows us to solve the first equation for \( y \):
\[
y = \frac{1}{b}(x' - ax)
\]
\[
\therefore y' = \frac{1}{b} x'' - \frac{a}{b} x'.
\]
The second equation tells us that \( y' = cx + dy \) so that equating the two yields
\[
\begin{align*}
    cx + dy &= \frac{1}{b} x'' - \frac{a}{b} x' \quad \longrightarrow \quad cx + \frac{d}{b} (x' - ax) = \frac{1}{b} x'' - \frac{a}{b} x' \\
    &\quad \longrightarrow \quad cbx + dx' - adx = x'' - ax' \\
    &\quad \longrightarrow \quad x'' - \text{trace}(A)x' + \det(A)x = 0.
\end{align*}
\]
This is a homogeneous, differential equation of a single variable. Solving this requires us to find \( r \) where
\[
r^2 - \text{trace}(A)r + \det(A) = 0.
\]
Note, however, that $\lambda$ is an eigenvalue iff $\lambda$ satisfies
\[\det(A - \lambda I) = 0 \iff \det\left(\begin{bmatrix} a - \lambda & b \\ c & d - \lambda \end{bmatrix}\right) = 0\]
\[\iff (a - \lambda)(d - \lambda) - bc = \lambda^2 - (a + d)\lambda + ad - bc = 0\]
\[\iff \lambda^2 - \text{trace}(A)\lambda + \det(A) = 0.\]
Hence the $\lambda$s we find will be eigenvalues. Since we assumed there were two distinct eigenvalues $\lambda_1$ and $\lambda_2$, this gives
\[x(t) = Ce^{\lambda_1 t} + De^{\lambda_2 t}\]
\[x'(t) = \lambda_1 Ce^{\lambda_1 t} + \lambda_2 De^{\lambda_2 t}\]
\[y(t) = \frac{1}{b}(x'(t) - ax(t))\]
\[= \frac{\lambda_1}{b} Ce^{\lambda_1 t} + \frac{\lambda_2}{b} De^{\lambda_2 t} - \frac{a}{b} Ce^{\lambda_1 t} - \frac{a}{b} De^{\lambda_2 t}\]
\[= \left(\frac{\lambda_1 - a}{b}\right) Ce^{\lambda_1 t} + \left(\frac{\lambda_2 - a}{b}\right) De^{\lambda_2 t}.\]
\[\therefore z(t) = \begin{bmatrix} x(t) \\ y(t) \end{bmatrix} = C \begin{bmatrix} 1 \\ (\lambda_1 - a)/b \end{bmatrix} e^{\lambda_1 t} + D \begin{bmatrix} 1 \\ (\lambda_2 - a)/b \end{bmatrix} e^{\lambda_2 t}.\]
So it suffices to show that
\[\begin{bmatrix} 1 \\ (\lambda_1 - a)/b \end{bmatrix}, \quad \text{and} \quad \begin{bmatrix} 1 \\ (\lambda_2 - a)/b \end{bmatrix}\]
are eigenvectors with eigenvalues $\lambda_1$ and $\lambda_2$ respectively. To show that this is the case, we merely multiply these vectors by $A$ on the left. Given that $\lambda^2 - (a + d)\lambda + (ad - bc) = 0$ for each $\lambda$, we get that
\[d\lambda - (ad - bc) = \lambda^2 - a\lambda:\]
\[\begin{bmatrix} a & b \\ c & d \end{bmatrix} \cdot \begin{bmatrix} 1 \\ (\lambda - a)/b \end{bmatrix} = \begin{bmatrix} a + (\lambda - a) \\ c + (d\lambda - da)/b \end{bmatrix} = \begin{bmatrix} \lambda \\ (d\lambda + cb - da)/b \end{bmatrix}\]
\[= \begin{bmatrix} \lambda \\ (\lambda^2 - a\lambda)/b \end{bmatrix} = \begin{bmatrix} 1 \\ (\lambda - a)/b \end{bmatrix}\]
Thus each vector is an eigenvector with the corresponding eigenvalue. This means that
\[z(t) = \begin{bmatrix} x(t) \\ y(t) \end{bmatrix} = C \begin{bmatrix} 1 \\ (\lambda_1 - a)/b \end{bmatrix} e^{\lambda_1 t} + D \begin{bmatrix} 1 \\ (\lambda_2 - a)/b \end{bmatrix} e^{\lambda_2 t}\]
\[= Cve^{\lambda_1 t} + Dwe^{\lambda_2 t}\]
where $v$ and $w$ are eigenvectors with eigenvalues $\lambda_1$ and $\lambda_2$ respectively.
Reference

Index

Abel’s theorem, 38
Absolute value, 4
Augmented matrix, 68, 74

Chain rule, 2, 18
Complex conjugate, 4, 69, 98
Complex number, 3, 4, 9, 10
Conjugate, see Complex conjugate
Critical point, see Solution, constant

Determinant, 70, 78
Difference equation, 29, 33, 34
Differential equation
  autonomous, 17, 26
  exact, 18, 19
  homogeneous, 37, 40, 41, 50
  homogeneous system, 83, 87, 100
  linear, 12, 16, 23, 37
  linear system, 83, 89, 101
  $n^{th}$-order, 12, 37, 55, 91
  non-homogeneous, 41, 50
  non-homogeneous system, 87, 99
  ordinary, 12
  partial, 12, 19
  separable, 12, 13, 17, 20, 21
  system, 83, 87, 88, 91
Direction field, 11

Eigenvalue, 72, 78
Eigenvalues, 85
Eigenvector, 72, 78, 86
Elementary row operation, 66–68
Euler’s formula, 4
Euler’s method, 30, 32
Euler’s method, improved, 31
Existence and uniqueness, 14

First-order, see Differential equation, $n^{th}$-order
  see also Difference equation

Gaussian elimination, 65, 67, 68, 74
Global truncation error, 30

Imaginary numbers, 4
Imaginary part, 4
improved Euler’s method, see Euler’s method, improved
Initial value, 12–14, 21
Integrating factor
  for exact equations, 19, 24
  for linear equations, 15, 20
Integration by parts, 3, 6

Invertible, see Multiplicative inverse

Jacobian matrix, 89, 103

Leading coefficient, 67
Linear combination, 37, 38
Linear dependence, 71, 72, 78, 85
Linear independence, 71, 96
Linearization, 89, 102
Local truncation error, 30, 31, 36

Magnitude, see Absolute value
Multiplicative inverse, 4, 65, 76

Polynomial, 5, 53
Product rule, 2, 3

Real part, 4
Recursion, 29
Reduced row echelon form, 66–68
Reduction of order, 39

Second-order, see Differential equation, $n^{th}$-order
Semistable, see Stability, semistable

Solution
  closed form, 29, 33
  constant, 17, 18, 26, 29
  equilibrium, see constant
  exact, see particular
  fundamental set, 38, 59, 85
  general, 1, 12, 20
  particular, 1, 16, 20

Stability
  asymptotically stable, 89, 90
  semistable, 18, 26
  stable, 18, 26, 29, 90
  unstable, 18, 26, 29, 90, 102

The chain rule, see Chain rule
The product rule, see Product rule
transpose, 69, 79

$u$-substitution, 3, 13

Undetermined coefficients
  for first-order equations, 16, 20
  for second-order equations, 42, 52

Unstable, see Stability, unstable

Variation of parameters, 42, 56

Wronskian, 37, 46