

MATH 320 NOTES, WEEK 15

5.4 Invariant Subspaces and Cayley-Hamilton theorem

The goal of this section is to prove the Cayley-Hamilton theorem:

Theorem 1. *Let $T : V \rightarrow V$ be a linear operator, V finite dimensional, and let $f(t)$ be the characteristic polynomial of T . Then $f(T) = T_0$ i.e. the zero linear transformation. In other words T is a root of its own characteristic polynomial.*

Here, if $f(t) = a_n t^n + a_{n-1} t^{n-1} + \dots + a_1 t + a_0$, plugging in T means the transformation

$$f(T) = a_n T^n + a_{n-1} T^{n-1} + \dots + a_1 T + a_0 I$$

Let us give some simple examples:

Example 1 The identity $I : F^3 \rightarrow F^3$ has characteristic polynomial $f(t) = (1-t)^3$. Then $f(I) = (I-I)^3 = T_0$.

Example 2 Let $A = \begin{pmatrix} 1 & 0 & 5 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix}$. Then the characteristic polynomial is

$$f(t) = (1-t)^2(2-t), \text{ and } f(A) = (A-I)^2(2I_3-A) = \begin{pmatrix} 0 & 0 & 5 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}^2 \begin{pmatrix} 1 & 0 & 5 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} = O.$$

We will prove the main theorem by using invariant subspaces and showing that if W is T -invariant, then the characteristic polynomial of $T \upharpoonright W$ divides the characteristic polynomial of T . So, let us recall the definition of a T -invariant space:

Definition 2. *Given a linear transformation $T : V \rightarrow V$, a subspace $W \subset V$ is called **T -invariant** if for all $x \in W$, $T(x) \in W$. For such a W , let $T_W : W \rightarrow W$ denote the linear transformation obtained by restricting T to W i.e. for all $x \in W$, $T_W(x) = T(x) \in W$.*

Examples:

- (1) $V, \{\vec{0}\}$,
- (2) $\ker(T), \text{ran}(T)$,

(3) E_λ for any eigenvalue λ for T .

Let us prove the last item: suppose that $v \in E_\lambda$. We have to show that $T(v) \in E_\lambda$. Denote $y = T(v)$ and compute

$$T(y) = T(T(v)) = T(\lambda v) = \lambda T(v) = \lambda y.$$

So, y is also an eigenvector for λ . Then $y = T(v) \in E_\lambda$ as desired.

Next we give another important example of an invariant subspace.

Lemma 3. *Suppose that $T : V \rightarrow V$ is a linear transformation, and let $x \in V$. Then*

$$W := \text{Span}(\{x, T(x), T^2(x), \dots\})$$

is a T -invariant subspace. Moreover, if Z is any other T -invariant subspace that contains x , then $W \subset Z$.

Proof. First we show that W is T -invariant: let $y \in W$. We have to show that $T(y) \in W$. Since $y \in W$, by definition, for some natural number n , $y = T^n(x)$. Then $T(y) = T^{n+1}(x) \in W$.

Now suppose that Z is another T -invariant subspace with $x \in Z$.

Claim 4. *For every $n \geq 1$, $T^n(x) \in Z$.*

Proof. For the base case $n = 1$, since $x \in Z$ and Z is T -invariant, it follows that $T(x) \in Z$.

For the inductive case, suppose that $T^n(x) \in Z$. Then again, by T -invariance, we have that $T^{n+1}(x) \in Z$. \square

By the claim, we get that $W \subset Z$. \square

W as above is called the **T -cyclic subspace of V generated by x** .

Example. Let $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be given by $T(\langle a, b, c \rangle) = \langle 2a, a + b, 0 \rangle$. Find the T -cyclic subspace of V generated by e_1 .

Solution:

- $T(e_1) = \langle 2, 1, 0 \rangle$,
- $T^2(e_1) = T(\langle 2, 1, 0 \rangle) = \langle 4, 3, 0 \rangle$, and so on

Note that $T^2(e_1)$ is a linear combination of $e_1, T(e_1)$. Similarly, for any n , $T^n(e_1) = \langle a_1, a_2, 0 \rangle$ for some a_1, a_2 , and so it is a linear combination of e_1 and $T(e_1)$. It follows, that the T -cyclic subspace of V generated by e_1 is $\text{Span}(\{e_1, T(e_1)\}) = \{\langle a_1, a_2, 0 \rangle \mid a_1, a_2 \in \mathbb{R}\} = \text{Span}(\{e_1, e_2\})$.

Our next lemma generalizes the above example:

Lemma 5. *Suppose that $T : V \rightarrow V$ is linear, let W be the T -invariant cyclic subspace generated by x (nonzero vector) with $\dim(W) = k$. Then $\{x, T(x), \dots, T^{k-1}(x)\}$ is a basis for W*

Proof. Let m be the largest such that $\alpha = \{x, T(x), \dots, T^{m-1}(x)\}$ is a linearly independent. Such m has to exist because W is finite dimensional. Then we have:

- $m \leq k$, since $\alpha \subset W$ and $\dim(W) = k$, and
- $T^m(x) \in \text{Span}(\alpha)$, by definition of m .

Let $Z = \text{Span}(\alpha)$. We claim that $Z = W$. We know that $Z \subset W$ because $\alpha \subset W$. For the other direction, by the second part of Lemma 3, it is enough to show that Z is T -invariant.

To that end, let $y \in Z$; write y as a linear combination of the vectors in α ,

$$y = a_1x + a_2T(x) + \dots + a_mT^{m-1}(x).$$

Compute

$$T(y) = T(a_1x + a_2T(x) + \dots + a_mT^{m-1}(x)) = a_1T(x) + a_2T^2(x) + \dots + a_mT^m(x).$$

This is a linear combination of vectors in α and $T^m(x)$. Since $T^m(x) \in \text{Span}(\alpha)$, we get $T(y) \in \text{Span}(\alpha) = Z$.

Then α is a basis for W , and so $m = |\alpha| = k$. □

Before we prove that the characteristic polynomial of T_W divides the characteristic polynomial of T where W is T -invariant, we need the following fact.

Fact 6. *Suppose we have an $n \times n$ matrix B of the form*

$$B = \begin{pmatrix} A & C \\ 0 & D \end{pmatrix},$$

Where A is a $k \times k$ matrix. Then $\det(A) \cdot \det(D)$

Proof. The proof is by induction on k , expanding along the first column. □

Lemma 7. *Suppose that $T : V \rightarrow V$ is linear, V finite dimensions, and W is a T -invariant subspace. Let $T_W : W \rightarrow W$ be the linear transformation obtained by T restricted to W . Then the characteristic polynomial of T_W divides the characteristic polynomial of T .*

Proof. Let $\alpha = \{v_1, \dots, v_k\}$ be a basis for W , and extend α to a basis $\beta = \{v_1, \dots, v_k, \dots, v_n\}$ for V . Let $A = [T_W]_\alpha$ and $B = [T]_\beta$. Then

$$B = \begin{pmatrix} A & C \\ 0 & D \end{pmatrix}$$

So,

$$(B - tI_n) = \begin{pmatrix} A - tI_k & C \\ 0 & D - tI_{n-k} \end{pmatrix}$$

Then $\det(B - tI_n) = \det(A - tI_k) \det(D - tI_{n-k})$. Since the characteristic polynomial of T_W is $\det(A - tI_k)$ and the characteristic polynomial of T is $\det(B - tI_n)$, the result follows. □

Lemma 8. *Suppose that $T : V \rightarrow V$ is linear, V finite dimensional, and let W be the T -invariant cyclic subspace generated by x (nonzero vector) with $\dim(W) = k$. Then there are unique scalars a_0, \dots, a_{k-1} , such that*

- $a_0x + a_1T(x) + \dots + a_{k-1}T^{k-1}(x) + T^k(x) = \vec{0}$, and

- the characteristic polynomial of T_W is

$$f(t) = (-1)^k(a_0 + a_1t + \dots a_{k-1}t^{k-1} + t^k).$$

Proof. Recall that we proved $\alpha = \{x, T(x), \dots, T^{k-1}(x)\}$ is a basis for W . That means that $T^k(x)$ is the span of these vectors and the linear combination is unique. So for some unique scalars a_0, \dots, a_{k-1} , we get:

$$a_0x + a_1T(x) + \dots + a_{k-1}T^{k-1}(x) + T^k(x) = \vec{0}.$$

Next we compute $[T_W]_\alpha$. We have:

- $T_W(x) = 0x + T(x) + 0\dots + 0$, so $[T_W(x)]_\alpha = \langle 0, 1, 0, \dots, 0 \rangle$,
- $T_W(T(x)) = T^2(x) = T_W^2(x)$, so $[T_W(T(x))]_\alpha = \langle 0, 0, 1, \dots, 0 \rangle$,
- ...
- $T_W(T^{k-1}(x)) = T^k(x) = -1(a_0x + a_1T(x) + \dots a_{k-1}T^{k-1}(x))$, and so $[T_W(T^{k-1}(x))]_\alpha = \langle -a_0, -a_1, \dots, -a_{k-1} \rangle$,

Then

$$[T_W]_\alpha = \begin{pmatrix} 0 & 0 & \dots & 0 & -a_1 \\ 1 & 0 & \dots & 0 & -a_2 \\ \vdots & & & & \\ 0 & 0 & \dots & 1 & -a_{k-1} \end{pmatrix}$$

One can verify that this matrix has characteristic polynomial

$$f(t) = (-1)^k(a_0 + a_1t + \dots a_{k-1}t^{k-1} + t^k)$$

□

We are finally ready to prove the Cayley-Hamilton theorem:

Proof of Theorem 1. We have to show that T “satisfies” its characteristic polynomial. So let $f(t)$ be the characteristic polynomial of T . Then we have to show that $f(T)$ is the zero linear transformation. So let $x \in V$. We have to show that $f(T)(x) = \vec{0}$.

Assume that x is nonzero (otherwise it’s clear). Let W be the T -cyclic invariant subspace generated by x ; $\dim(W) = k$. By the above lemma, fix coefficients a_0, \dots, a_k , such that

- $a_0x + a_1T(x) + \dots a_{k-1}T^{k-1}(x) + T^k(x) = \vec{0}$, and
- the characteristic polynomial of T_W is

$$g(t) = (-1)^k(a_0 + a_1t + \dots a_{k-1}t^{k-1} + t^k).$$

Then,

$$\begin{aligned} g(T)(x) &= (-1)^k(a_0I + a_1T + \dots a_{k-1}T^{k-1} + T^k)(x) = \\ &(-1)^k(a_0x + a_1T(x) + \dots a_{k-1}T^{k-1}(x) + T^k(x)) = \vec{0}. \end{aligned}$$

And since g divides f , we get that $f(T)(x) = 0$.

□

6.1, 6.2 Inner Product Spaces and ONB

Let $c \in F$, we will use \bar{c} to denote complex conjugation. I.e. if $c = a + bi$, then $\bar{c} = a - bi$. If $c \in \mathbb{R}$, then $\bar{c} = c$.

Definition 9. Let V be a vector space over F . V is an **inner product space**, if we can define an inner product function on pairs of vectors to values in F ,

$$(x, y) \mapsto \langle x, y \rangle \in F$$

with the following properties:

- (1) $\langle x + z, y \rangle = \langle x, y \rangle + \langle z, y \rangle$
- (2) $\langle cx, y \rangle = c\langle x, y \rangle$,
- (3) $\langle x, y \rangle = \overline{\langle y, x \rangle}$,
- (4) if $x \neq \vec{0}$, $\langle x, x \rangle > 0$,

Definition 10. Let V be an inner product space. For $x \in V$, define the **norm** of x ,

$$\|x\| = \sqrt{\langle x, x \rangle}.$$

Lemma 11. (Properties of inner product spaces) Let V be an inner product space, and $x, y, z \in V$, $c \in F$. Then

- (1) $\langle x, y + z \rangle = \langle x, y \rangle + \langle x, z \rangle$
- (2) $\langle x, cy \rangle = \bar{c}\langle x, y \rangle$,
- (3) $\langle x, \vec{0} \rangle = \langle \vec{0}, x \rangle = 0$,
- (4) $\langle x, x \rangle = 0$ iff if $x = \vec{0}$,
- (5) if $\langle x, y \rangle = \langle x, z \rangle$ for all x , then $y = z$.

Lemma 12. (Properties of norms) Let V be an inner product space over F , and $x, y, z \in V$, $c \in F$. Then

- (1) $\|cx\| = |c| \cdot \|x\|$,
- (2) $\|x\| \geq 0$, $\|x\| = 0$ iff if $x = \vec{0}$,
- (3) (Cauchy-Schwarz inequality) $|\langle x, y \rangle| \leq \|x\| \cdot \|y\|$,
- (4) (Triangle inequality) $\|x + y\| \leq \|x\| + \|y\|$.

Examples:

- (1) The usual dot product for \mathbb{R}^n over \mathbb{R} : $\langle x, y \rangle = \sum_{1 \leq i \leq n} x_i y_i$;
- (2) The conjugate dot product for \mathbb{C}^n over \mathbb{C} : $\langle x, y \rangle = \sum_{1 \leq i \leq n} x_i \bar{y}_i$;

Definition 13. Let V be an inner product space. Two vectors x, y in V are called **orthogonal** (or perpendicular) if $\langle x, y \rangle = 0$.

A vector x in V is called **normal** if $\|x\| = 1$.

A set S is called **orthonormal** if any two vectors from S are orthogonal and each $x \in S$ is normal.

Lemma 14. If S is a set of pairwise orthogonal nonzero vectors, then S is linearly independent. So, orthonormal sets are linearly independent.

Proof. Suppose

$$a_1v_1 + \dots + a_kv_k = \vec{0}$$

for vectors v_1, \dots, v_k in S . Suppose for contradiction some of the a_i 's are non zero. By rearranging if necessary, we may assume that $a_1 \neq 0$. Then

$$0 = \langle a_1v_1 + \dots + a_kv_k, v_1 \rangle = a_1\langle v_1, v_1 \rangle + a_2\langle v_2, v_1 \rangle + \dots + a_k\langle v_k, v_1 \rangle = a_1\langle v_1, v_1 \rangle.$$

By assumption $v_1 \neq \vec{0}$, and so by the properties of inner products $\langle v_1, v_1 \rangle > 0$, so $a_1 = 0$. Contradiction \square

Using a process called Gram-Schmidt orthogonalization, we can obtain an orthonormal set from any basis, and get the following theorem:

Theorem 15. *Let V be an inner product space of dimension $n > 0$. Then there is an orthonormal basis for V (ONB).*

Example: the standard basis is an ONB for F^n .

The next lemma illustrates the usefulness of ONBs, in the sense that we can determine in advance the coefficients of linear combinations and matrix representations.

Lemma 16. *If $\beta := \{v_1, \dots, v_n\}$ is an ONB for V and $T : V \rightarrow V$ is linear, then*

(1) *For every $x \in V$,*

$$x = \langle x, v_1 \rangle v_1 + \langle x, v_2 \rangle v_2 + \dots + \langle x, v_n \rangle v_n$$

(2) *If $A = [T]_\beta$, then the (i, j) -th entry of the matrix is $A_{ij} = \langle T(v_j), v_i \rangle$.*

Proof. We show the first part. The second is left as an exercise. Fix $x \in V$. Say $x = a_1v_1 + \dots + a_nv_n$. Then for each $1 \leq i \leq n$,

$$\langle x, v_i \rangle = \langle a_1v_1 + \dots + a_nv_n, v_i \rangle = \sum_k a_k \langle v_k, v_i \rangle = a_i \langle v_i, v_i \rangle = a_i.$$

The above is since for $k \neq i$, $\langle v_k, v_i \rangle = 0$ and $\langle v_i, v_i \rangle = 1$. \square

Definition 17. *Let $S \subset V$ be nonempty, V an inner product space. The orthogonal complement of S is*

$$S^\perp = \{x \in V \mid \langle x, y \rangle = 0 \text{ for all } y \in S\}$$

Fact 18. *For any nonempty set S , S^\perp is a subspace.*

A couple of examples:

- $\{\vec{0}\}^\perp = V$; $V^\perp = \{\vec{0}\}$,
- in F^3 , $\{e_1\}^\perp = \text{Span}(\{e_2, e_3\})$.

6.3, 6.4 Adjoint operators and the Spectral theorem

Let us first look at matrices:

Definition 19. Let $A \in M_{m \times n}(F)$. The **conjugate transpose** of A is the $n \times m$ matrix A^* given by setting $A_{ij}^* = \bar{A}_{ji}$ i.e. we take the conjugate of each entry and transpose it. We also write $A^* = \bar{A}^t$.

We can also define the adjoint for a linear transformations in general.

Theorem 20. Let $T : V \rightarrow V$ be a linear operator on a finite dimensional inner product space. Then there is a linear transformation $T^* : V \rightarrow V$, such that for all $x, y \in V$,

$$\langle T(x), y \rangle = \langle x, T^*(y) \rangle.$$

T^* is called the **adjoint** of T .

Moreover, if β is an ONB, then $[T^*]_\beta = [T]_\beta^*$.

The following is a key lemma:

Theorem 21. (Schur) Suppose that the characteristic polynomial of T splits. Then there is an ONB β such that $[T]_\beta$ is upper triangular.

Proof. (Outline) By induction on $\dim(V)$. The idea is to pick one eigenvalue (it exists since the characteristic polynomial splits), then a corresponding eigenvector z . Then prove that $W := \text{Span}(z)^\perp$ is T -invariant and apply the inductive hypothesis to T_W . □

Next we give a condition for a diagonalizability over \mathbb{R} .

Definition 22. A linear transformation $T : V \rightarrow V$ is **self-adjoint (Hermitian)** if $T = T^*$. Similarly, a matrix A is **self-adjoint (Hermitian)** if $A = A^*$.

Lemma 23. If T is self adjoint, then every eigenvalue is real and the characteristic polynomial splits over \mathbb{R} .

Proof. Suppose that λ is an eigenvalue with eigenvector x . Then $\lambda \langle x, x \rangle = \langle \lambda x, x \rangle = \langle T(x), x \rangle = \langle x, T^*(x) \rangle = \langle x, T(x) \rangle = \langle x, \lambda x \rangle = \bar{\lambda} \langle x, x \rangle$. Then $\lambda = \bar{\lambda}$, and so it is real.

Let $f(t)$ be the characteristic polynomial. Then $f(t)$ splits over \mathbb{C} . But since every eigenvalue is real, every root of $f(t)$ is real, and so it must split over \mathbb{R} . □

Theorem 24. (Spectral theorems for real spaces) Suppose that $T : V \rightarrow V$ is linear, V is a finite dimensional real inner product space. Then T is self adjoint iff it has an ONB of eigenvectors.

Proof. By the above lemma the characteristic polynomial of T splits. So by theorem 21, let β be ONB such that $A := [T]_\beta$ is upper triangular. Then $[T^*]_\beta = A^*$ has to be lower triangular. But since $A = A^*$, it follows that A is diagonal. So β is a basis of eigenvectors and T is diagonalizable. □

In other words, symmetric real matrices are diagonalizable.

Below we briefly state (without proofs), the case of complex spaces.

Definition 25. $T : V \rightarrow V$ is **normal** if $TT^* = T^*T$.

Theorem 26. (*Spectral theorem for complex spaces*) Suppose that $T : V \rightarrow V$ is linear, V is a finite dimensional complex inner product space. Then T is normal iff it has an ONB of eigenvectors.