“Let us imagine the following example: A writes series of numbers down; B watches him and tries to find a law for the sequence of numbers. If he succeeds he exclaims: “Now I can go on!” - So this capacity, this understanding, is something that makes its appearance in a moment…”

-Ludwig Wittgenstein, *Philosophical Investigations 1*, §151. [emphasis added]

**Seven-Year-Old Gauss**

We all know how, allegedly, young Carl Friedrich Gauss summed-up the first 100 natural numbers, by using an “ingenious” trick. But this trick would fail miserably if he had, instead, to sum the first 100 perfect squares. A much better way would have been for him to compute the first few terms of the sequence $a(n) := 1 + 2 + \ldots + n$:

- $a(0) = 0$
- $a(1) = 1$
- $a(2) = 3$
- $a(3) = 6$
- $a(4) = 10$

and then using his genius for detecting patterns, notice that

- $a(0) = (0 \cdot 1)/2$
- $a(1) = (1 \cdot 2)/2$
- $a(2) = (2 \cdot 3)/2$
- $a(3) = (3 \cdot 4)/2$
- $a(4) = (4 \cdot 5)/2$

then conjecture that $a(n) = n(n + 1)/2$, for every $n$. Then he could have tested it for a few more cases, and confirmed that the conjecture is correct for $n = 5$ and $n = 6$, and deduced, that $a(100) = (100 \cdot 101)/2 = 5050$. See [Z1]. [In fact checking it for $n = 0, 1, 2$ constitutes a fully rigorous proof of the identity, since both sides are polynomials of degree 2.]

In this case detecting a pattern was easy enough, we will soon see examples where detecting patterns is much harder, and would need computers.

**[Physical] Induction vs. Deduction**

Everyone knows that, unlike the physical sciences, that use inductive logic, and are empirical, mathematics is a deductive science, where one proceeds from axioms, and, step by step, using solid logic, proceeds to prove theorems. Of course, this is a gross oversimplification, and already the great James Joseph Sylvester took issue with this conventional wisdom uttered by Thomas Huxley who claimed that

“Mathematics is that study which knows nothing of experiment, nothing of induction, nothing of causation”,

and Sylvester ([Sy]) retorted that mathematics is
“unceasingly calling forth the faculties of observation and comparison, and one of its principal weapons is induction, that it has frequent recourse to experimental trial and verification ...”

And indeed, Sylvester was right that in that part of mathematics that philosophers of science call context of discovery, mathematicians use experiments and induction (in the sense of physical induction, not to be confused with \[\text{complete}]\ mathematical induction that is a fully deductive process). But in the context of justification, i.e. when presenting the proof, deductive logic rules.

**Evolution Did not prepare us for reasoning logically and for rigorous proofs**

In a beautiful recent masterpiece ([Ar]), Zvi Artstein claims, very convincingly, that the reason most people (in fact, all of them, including us, professional mathematicians!) find math so hard, is that biological evolution did not prepare us to the rigid discipline of formal mathematical and logical reasoning. In order to survive in the jungle, we had to use informal, intuitive, ‘Bayesian’ ‘logic’, if you would call it logic at all.

**Mathematical Cultural Evolution Did not prepare us to Use Computers Optimally**

But in spite of that, Mathematics blossomed, and has come a long way, both as queen and servant of the physical sciences. For more than two millenia, the cultural evolution of mathematics, and the notions of axiomatic method and rigorous proof ruled. But neither Euclid, nor Gauss, not even Ramanujan, knew about the new messiah, the powerful electronic computer, that would revolutionize both the discovery and the justification of mathematical knowledge, and would (soon!) turn mathematics into an empirical science, just like physics, chemistry, and biology, but dealing with mathematical entities (like numbers, equations, groups etc.) rather than with electrons and stars, (or acids and bases, or cells and genes etc.), and we would soon abandon our fanatical insistence on rigorous proof, and very soon semi-rigorous proofs ([Z2]) would be fully acceptable, and soon after, completely non-rigorous proofs!

We will describe below, as case studies, many examples where ‘rigorous proofs’ are possible, but not worth the trouble, since they are so computer-time and computer-memory consuming. In such cases, semi-rigorous, and even non-rigorous proofs, are good enough!

**Revisiting an Old Chestnut: In how many ways can a gambler win n dollars and lose n dollars, and never be in debt?**

Consider a gambler who tosses a coin \[2n\] times, and if it comes out Heads, wins a dollar, and if it comes out Tails, loses a dollar. He is kicked out as soon as he is in the red, i.e. has negative capital. In how many ways can he survive to \[2n\] rounds, but at the end break even?

For example, denoting a win by \(W\) and a loss by \(L\), the following is the set of such ‘good’ gambling histories, with \(n = 3\):
Note, for example, that \( \overline{WLLWL} \) is not allowed, since after the third round, the gambler owes a dollar. How to count these debt-less sequences of Wins and Losses?

Let’s first consider the easier problem where the gambler is allowed to be in debt. The number of ways of doing it is \( \binom{2n}{n} \) (pronounced “\( 2n \) choose \( n \)”), since they were \( n \) wins and \( n \) losses, and Lady Luck had to choose out of the total number of \( 2n \) coin-tosses, which of them are wins (and the remaining ones are losses).

Now what is the number of such histories where he was never in debt? A classical, very famous result in enumerative combinatorics (see for example, [Ai], p. 98) says that this number is given by the ubiquitous Catalan numbers \([\text{Sl}]\) \( \frac{1}{n+1} \binom{2n}{n} \). It follows that, assuming a fair coin, that if the gambler is allowed to get credit, and tossed the coin \( 2n \) times, and broke even at the end, the probability that he never had to borrow is exactly \( \frac{1}{n+1} \).

We know at least ten proofs of this result, some more elegant than others, but here we will present yet another, computer-assisted proof, that is very possibly the ugliest. Its advantage is that the same method can prove, at least in principle, but very often also in practice, any problem of this type, where instead of the ‘atomic’ stakes being taken from the set \( \{-1, 1\} \), it is taken from any given (finite) set of integers, \( S \). For example, \( S = \{-1, -2, 3\} \) (see below). The same method extends to many (but not all) ‘walks’ in two dimensions, where at each round he gets \( a \) dollars and \( b \) euros where \([a, b]\) belongs to some given finite set of pairs \( S \) (and where ‘getting’ a negative amount, means paying), and both his number of dollars and number of euros must always be non-negative. The method extends to three and more currencies, but then the computations take way too long for today’s computers.

In order to illustrate our ‘empirical’, guess-and-check method, we will, as promised, use it to give the 101-th proof of the above-mentioned famous result that the number of gambling histories with \( n \) wins and \( n \) losses, and never being in debt, is indeed given by the Catalan number \( \frac{1}{n+1} \binom{2n}{n} = \frac{(2n)!}{n!(n+1)!} \). While, like all our proofs, it was computer-generated, it is sufficiently simple to be followed and understood by humans with some patience. Even the discovery of the proof, in this simple example, could be done by hand, and we will do it. This is for purely pedagogical reasons, to make the method understandable to humans, so that they can understand how our silicon colleagues can handle much more complicated cases.

We will reprove the equivalent statement

**Theorem:** Let \( a(n) \) be the number of \( n \)-long sequences \( w_1 \ldots w_n \) with \( w_i \in \{-1, 1\} \), such that \( \sum_{j=1}^{i} w_j \geq 0 \) for \( 1 \leq i \leq n - 1 \) and \( \sum_{j=1}^{n} w_j = 0 \). The generating function

\[
f(t) := \sum_{n=0}^{\infty} a(n)t^n
\]
is given, explicitly, by
\[ f(t) = \frac{1 - \sqrt{1 - 4t^2}}{2t^2}. \]

[Note that if \( n \) is odd, \( a(n) = 0 \), of course, and extracting the coefficient of \( t^{2n} \) from the right side gives \(-\frac{1}{2}(\frac{3}{2})(-4)^{n+1} \) that simplifies to the Catalan number \( \frac{1}{n+1} \binom{2n}{n} \).]

**Proof:** Consider, more generally, all sequences \( w_1 \ldots w_n \) with \( w_i \in \{-1, 1\} \), with the property that the partial sums, \( \sum_{j=1}^{n} w_j \geq 0 \), are all non-negative, but without the restriction that the total sum, \( \sum_{j=1}^{n} w_j \), equals 0. Let the *weight* of such a sequence of length \( n \), in \( \{-1, 1\} \), be \( t^n x^{TotalSum} \), in other words:
\[ Weight(w_1 w_2 \ldots w_n) := t^n x^{w_1 + \ldots + w_n}. \]

For example
\[
\begin{align*}
Weight(EmptySequence) &= t^0 x^0 = 1, \\
Weight([1, 1, -1, 1, -1]) &= t^5 x^1 = t^5 x, \\
Weight([1, 1, -1, 1, 1, 1, 1, 1, -1]) &= t^{10} x^2.
\end{align*}
\]

Let \( \mathcal{W} \) be the set of such sequences (including the empty sequence of length 0, whose weight is \( t^0 x^0 = 1 \)), and let \( F(t, x) \) be the sum of all the weights, a certain *formal power series*, in the variable \( t \), whose coefficients are polynomials in \( x \). Note that the coefficient of \( t^n x^i \) is the exact number of ‘good’ sequences \( w_1 \ldots w_n \) whose entries are taken from \( \{-1, 1\} \) where all the partial sums are non-negative, and the total sum is \( i \). In particular, Note that \( f(t) \), our object of desire, is simply \( F(t, 0) \).

We claim that \( F(t, x) \) satisfies the *functional equation*
\[ F(t, x) = 1 + txF(t, x) + tx^{-1}(F(t, x) - F(t, 0)) \quad (FunctionalEquation) \]

Indeed consider any \( w_1 \ldots w_n \in \mathcal{W} \).

**Case 1:** It is the empty sequence, (i.e. \( n = 0 \)), whose weight is 1.

**Case 2:** It ends with a 1, i.e. \( w_n = 1 \), then \( w_1 \ldots w_{n-1} \) is a legal sequence, and all members of \( \mathcal{W} \) of length \( n \) that end with 1 are obtained by taking sequences of length \( n - 1 \) and appending 1 to them. Hence the sum of the weights of the members of \( \mathcal{W} \) that end with 1 is
\[ txF(t, x) \quad , \]

* A formal power series is an infinite power series of the form \( \sum_{n=0}^{\infty} a_n t^n \) where we do not worry about (analytic) convergence, for example, \( \sum_{n=0}^{\infty} n! t^n \) is perfectly OK. One can redo most of classical analysis via the algebra of formal power series, that ironically, in spite of the name ‘formal’, is much more rigorous than the traditional analysis we learn in college, since it only uses finitistic notions.
since appending the 1 increases both the t-count and the x-count by 1.

**Case 3:** It ends with a ‘−1’, i.e. \( w_n = -1 \), then \( w_1 \cdots w_{n-1} \) is a legal sequence but, in addition, has the property that \( \sum_{i=1}^{n-1} w_i > 0 \), so the set of sequences of length \( n \) that end with a ‘−1’ are obtained by taking walks of length \( n - 1 \), except those that add up to 0, and appending ‘−1’ to them. In other words, we have to exclude those sequence of length \( n - 1 \) that sum-up to zero, whose total weight is \( F(t,0) \). Hence, the sum of all the weights of the members of the set of good sequences that end with ‘−1’ is

\[
t x^{-1} (F(t,x) - F(t,0))
\]

since appending the ‘−1’ increases the t-count (the length of the walk) by 1, but decreases the x-count by 1, and \( F(t,x) - F(t,0) \) is the sum of the weights of walks whose total sum is strictly positive.

This concludes the proof of the Functional Equation. Note that by iterating the mapping

\[
f(t,x) \rightarrow 1 + t x f(t,x) + t x^{-1} (f(t,x) - f(t,0))
\]

\( n \) times, starting with \( f(t,x) = 1 \), we get an efficient way to generate the first \( n+1 \) coefficients, in \( t \), of the formal power series \( F(t,x) \).

We now pull out of a hat, the explicit expression

\[
G(t,x) := \frac{1 - 2 x t - \sqrt{1 - 4 t^2}}{2 t (-x + t + t x^2)}
\]

and claim that \( F(t,x) = G(t,x) \). Indeed, dear readers, we are sure that, with some patience, and a bit of masochism, you would be able to verify, by hand, the purely routine, high-school algebra, calculation that if one replaces \( F(t,x) \) by \( G(t,x) \) in Eq. \( \text{(FunctionalEquation)} \), the two sides match. In other words, please check that

\[
G(t,x) - (1 + t x G(t,x) + t x^{-1} (G(t,x) - G(t,0))) \equiv 0.
\]

\( \text{(FunctionalEquation')} \)

If however, you are too lazy, you are welcome to ‘cheat’ and copy-and-paste the following Maple code to a Maple session:

```maple
G:=t,x->(1-2*x*t-sqrt(1-4*t**2))/(2*t*(-x+t+t*x**2));
simplify(G(t,x)-1-t*x*G(t,x)-t/x*(G(t,x)-G(t,0)));
```

and get 0.

There is one sticky point remaining. It is not immediately obvious that the \( G(t,x) \) is a formal power series in \( t \) with coefficients that are polynomials in \( x \), but the readers are welcome to use the good old formula for solving a quadratic equation \((-b \pm \sqrt{b^2 - 4a c})/2a\), to check that \( G(t,x) \) is a solution of the quadratic equation

\[
t (-x + t + t x^2) G(t,x)^2 + (-1 + 2 x t) G(t,x) + 1 = 0.
\]

\( \text{(AlgebraicEquation)} \)
This can be rewritten as
\[ G(t, x) = 1 + 2xt G(t, x) + t \left(-x + t + tx^2\right) G(t, x)^2 \]  \hspace{1cm} (AlgebraicEquation')
that manifestly shows that \( G(t, x) \) is indeed a formal power series in \( t \) with coefficients that are polynomials in \( x \), and the first \( n \) terms in its Maclaurin expansion can be gotten by starting with \( g(t, x) = 1 \) and applying the mapping
\[ g(t, x) \rightarrow 1 + 2xtg(t, x) + t \left(-x + t + tx^2\right) g(t, x)^2 \]  \hspace{1cm} (AlgebraicEquation')
n times.

So we have proved that \( F(t, x) \), the weight-enumerator (alias generating function) for our walks, is given by the explicit expression \( G(t, x) \). Finally to prove the theorem, just plug-in \( x = 0 \) and get
\[ f(t) = F(t, 0) = G(t, 0) = \frac{1 - 2 \cdot 0 \cdot t - \sqrt{1 - 4t^2}}{2t (-0 + t + t \cdot 0^2)} = \frac{1 - \sqrt{1 - 4t^2}}{2t^2} \] .
QED!

Secrets from the Kitchen

The above proof is the kind of proof that G.H. Hardy called, derogatorily, essentially verification. It pulled, out of the hat, the explicit expression \( G(t, x) \) and then used uniqueness to prove that \( F(t, x) = G(t, x) \). How can me come-up with \( G(t, x) \)?

By guessing, of course! If we have a premonition that \( F(t, x) \) satisfies, in addition to the natural functional equation, that was derived by combinatorial reasoning, also an algebraic equation of the form
\[ A_0(t, x) + A_1(t, x) F(t, x) + A_2(t, x) F(t, x)^2 = 0 \]  \hspace{1cm} (AlgebraicEquation)
for some polynomials \( A_0(t, x), A_1(t, x), A_2(t, x) \) to be determined, one can first crank-out the first, say, 20 coefficients, in \( t \), of \( F(t, x) \) and use linear algebra to guess the coefficients. But there is a much simpler way, that in this, toy example, can even be done by hand.

The functional equation \( (FunctionalEquation) \), can be rewritten (recall that \( F(t, 0) = f(t) \))
\[ F(t, x) = \frac{1 - tx^{-1}f(t)}{1 - tx - tx^{-1}} \]
so let’s try to guess a quadratic equation satisfied by \( f(t) \). Since \( f(t) \) only contains even powers, let’s consider instead \( h(t) := f(\sqrt{t}) \), and optimistically look for undetermined coefficients, \( c_1, c_2, c_3, c_4, c_5, c_6 \), such that
\[ (c_1 + c_2t) + (c_3 + c_4t)h(t) + (c_5 + c_6t)h(t)^2 \equiv 0 \] .

Less wastefully, knowing the combinatorial origin of \( f(t) \) (and hence \( h(t) \)), it is reasonable to look for numbers \( d_1, d_2, d_3 \) such that
\[ h(t) = 1 + d_1t h(t) + (d_2 + d_3t) h(t)^2 \]  \hspace{1cm} (Hope)
or equivalently
\[ h(t) - 1 - d_1 t h(t) - (d_2 + d_3 t) h(t)^2 \equiv 0 \quad , \tag{Hope'} \]
for some numbers, \(d_1, d_2, d_3\) that are yet to be determined. Of course, a priori, there is no guarantee that we would be successful, but let’s try!

It is easy to crank out, either using \((FunctionalEquation)\), or even by direct counting, the first few terms of \(h(t)\) (alias \(f(\sqrt{t})\))
\[ h(t) = 1 + t + 2t^2 + 5t^3 + 14t^4 + 42t^5 + O(t^6) \quad . \]
Now, plug-it-in into \((Hope')\):
\[ (1 + t + 2t^2 + 5t^3 + 14t^4 + 42t^5 + O(t^6)) - 1 - d_1 t (1 + t + 2t^2 + 5t^3 + 14t^4 + 42t^5 + O(t^6)) \]
\[ -(d_2 + d_3 t) (1 + t + 2t^2 + 5t^3 + 14t^4 + 42t^5 + O(t^6))^2 \equiv 0 \quad . \tag{Hope''} \]
Expanding, and collecting coefficients of \(t^i\) for \(i = 1\) to \(i = 5\) yields a system of five linear equations for the three unknowns \(d_1, d_2, d_3\), whose unique solution is \(d_1 = 0, d_2 = 0, d_3 = 1\). This leads us to conjecture that \(h(t)\) satisfies the algebraic equation
\[ h(t) = 1 + t h(t)^2 \quad , \tag{Yea!} \]
Replacing \(h(t^2)\) by \(f(t)\), we have guessed that \(f(t)\) is a solution of the quadratic equation
\[ f(t) = 1 + t^2 f(t)^2 \quad . \]
That is equivalent to the statement of the theorem, and that lead to the conjectured expression for \(F(t, x)\) (i.e. the expression \(G(t, x)\) above), that before we “pulled out of the hat”, and lead to our essentially verification proof.

What is an Answer?

In a classical paper ([W], see also [Z3]), guru Herb Wilf, defined what is a satisfactory answer to a combinatorial enumeration problem that, typically, asks to enumerate a sequence of sets parametrized by one or more discrete parameters. Obviously the answers \(2^n\) for the number of subsets of a set of \(n\) elements, and \(n!\) for the number of permutations of \(n\) objects seem satisfactory. Traditionally, one wanted an explicit ‘formula’ like in the above two examples. But what is a formula?, it is an algorithm for getting the answer, and some algorithms are better than others. In fact, there is always a ‘formula’ for enumerating any set,
\[ |A(n)| = \sum_{a \in A(n)} 1 \quad , \]
just generate all the elements of \(A(n)\), and then count them. Most combinatorial sets have exponential (or larger) sizes, so this ‘formula’ is (usually) useless! By using the modern computer-science
dichotomy of polynomial vs. exponential growth, Wilf suggested that a good answer is a polynomial time algorithm.

Going back to our toy example, the functional equation \( \text{(FunctionalEquation)} \) is already a good answer! It is easy to see that it requires cubic time \( (O(n^3)) \) and quadratic memory \( (O(n^2)) \). On the other hand, the ‘answers’ that we found, the explicit expression for the generating function, \( f(t) = \frac{1-\sqrt{1-4t}}{2t} \) is better, since it implies a quadratic in \( N \) algorithm for generating the first \( N \) terms, and the explicit expression, \( a(2n) = \frac{(2n)!}{(n!(n+1)!)} \) is yet better.

In our toy example, we found that \( h(t) = f(\sqrt{t}) \) satisfies the algebraic equation

\[ th(t)^2 - h(t) + 1 = 0 \ , \]

that turned out (in this simple case) to be quadratic, and hence solvable by radicals. Many combinatorial sequences \( c(n) \) have the property that their generating functions \( C(t) := \sum_{n=0}^{\infty} c(n)t^n \) satisfy algebraic equations of the form

\[ \sum_{i=0}^{d} p_i(t)C(t)^i = 0 \ , \]

(\( \text{where } p_i(t) \text{ are some polynomials of } t \)), but not necessarily with \( d \leq 4 \), so generally they are not ‘solvable by radicals’, but who cares? Just giving the minimal algebraic equation satisfied by the generating function \( C(t) \) of our desired sequence is a good answer in the sense of Wilf.

Sequences whose generating functions satisfy such algebraic equations are called algebraic, and it is well-known ([St][Z4]) that they form an algebra, and in particular it is decidable whether \( A = B \), so in order to prove that two sequences are identically equal, it suffices to check them for finitely many cases, from \( n = 0 \) to \( n = N_0 \), say, where \( N_0 \) can be (usually) easily found a priori.

It is easy to see that sequences whose generating functions satisfy algebraic equations satisfy a non-linear recurrence with constant coefficients, alas, requiring to remember all the past values. For example for the Calalan numbers (the coefficients of \( h(t) \)), we have the non-linear recurrence

\[ a(n) = \sum_{i=0}^{n-1} a(i)a(n-1-i) \ , \quad a(0) = 1 \ . \]

But in this toy example, the sequence itself, i.e. the number of good gambling histories, have an explicit, ‘closed-form’ answer, that is even better from the Wilfian point of view. Namely, we have

\[ a(n) = \frac{(2n)!}{n!(n+1)!} \ , \]

Since \( a(n+1)/a(n) = \frac{4n+2}{n+2} \), an equivalent description of the sequence \( a(n) \) is as the unique solution of the linear recurrence equation

\[ (n+2)a(n+1) - (4n+2)a(n) = 0 \ , \quad a(0) = 1 \ . \]
This is an example of yet another important ansatz, called ‘P-recursive’, or ‘discrete holonomic’. A P-recursive sequence \( \{c(n)\} \) is uniquely determined by a homogeneous linear recurrence equation with polynomial coefficients

\[
\sum_{i=0}^{d} q_i(n)c(n+i) = 0 ,
\]

for some given polynomials \( q_0(n), \ldots, q_d(n) \), together with the \( d \) initial values \( c(0), \ldots, c(d-1) \).

It is well-known ([St]) that this class, too, is an algebra, i.e. the sum and product of \( P \)-finite sequences are again \( P \)-finite sequences, and it is always possible to decide whether \( A = B \). Of course, usually the recurrence is not first-order, like in this case, but it is a finite and efficient description of the sequence, and once we know the recurrence, and the initial conditions, it is possible to compute its members in linear time and constant memory, since we can ‘forget’ about previous values, and only need to retain a ‘window’ as large as the order, \( d \), of the recurrence.

Equivalently, a sequence, \( c(n) \) is \( P \)-recursive if and only if its generating function, \( C(t) = \sum_{n=0}^{\infty} c(n)t^n \) satisfies a linear differential equation with polynomial coefficients

\[
\left( \sum_{i=0}^{L} r_i(t) \frac{d^i}{dt^i} \right) C(t) = 0 ,
\]

with the appropriate \( L \) initial conditions at 0. Such power series are called ([St]) \( D \)-finite, or simply (continuous) holonomic. The set of \( D \)-finite power series is also easily seen to be an algebra.

It is known ([St]), that every algebraic formal power series is always \( D \)-finite, and hence its sequence of coefficients is \( P \)-recursive. The converse is false! For example the \( P \)-recursive sequence \( n! \) can be easily seen not to have an algebraic generating function. The algebraic and holonomic ansatizes are nicely implemented in the Maple package \texttt{gfun} described in [SZ].

**Back to Gambling**

Suppose that the gambler has a three-sided coin, marked with \( \{-1, -2, 3\} \). How many gambling histories are there of length \( n \) where (i) the gambler broke even but was never in debt (ii) he was never in debt but not necessarily broke even at the end. Let’s call the first sequence \( a(n) \) and the second \( b(n) \).

The second-named author discovered the following

**Theorem:** Let \( b(n) \) be the number of sequences of length \( n \), \( w_1 \ldots w_n \) with \( w_i \in \{-1, -2, 3\} \) such that \( \sum_{j=1}^{i} w_j \geq 0 \) (for \( 1 \leq i \leq n \)), and let \( a(n) \) be the number of such sequences where in addition \( \sum_{j=1}^{n} w_j = 0 \).

The sequence \( a(n) \) satisfies a linear recurrence equation with (very complicated!) polynomial coefficients of order twenty that may be viewed in

http://www.math.rutgers.edu/~zeilberg/tokhniot/oW1D7
and $b(n)$ satisfies a linear recurrence equation with (equally complicated) polynomial coefficients of order twenty one that may be viewed in

http://www.math.rutgers.edu/~zeilberg/tokhniot/oW1D8

Let’s describe how we discovered this theorem, by treating the general case.

**The General Case**

Suppose you are given any (finite) set of integers $S$, and let $F(t, x)$ be the weight enumerator according to the same weight, $\text{Weight}(w_1 \ldots w_n) := t^n x^{w_1 + \ldots + w_n}$, of the set of sequences in $S^n$ whose partial sums are always non-negative (equivalently, the number of one-dimensional walks starting at 0, and never going to the negatives), then the same logic as before shows that $F(t, x)$ satisfies the functional equation

$$F(t, x) = 1 + t \sum_{s \in S, s \geq 0} x^s F(t, x) + t \sum_{s \in S, s < 0} x^s \left( F(t, x) - \frac{s-1}{x} \cdot \text{Coeff} x^s F(t, x) \right).$$

(GeneralFunctionalEquation)

For example, if $S = \{-1, -2, 3\}$, the functional equation is

$$F(t, x) = 1 + tx^3 F(t, x) + tx^{-1}(F(t, x) - F(t, 0)) + tx^{-2}(F(t, x) - F(t, 0) - x \cdot \text{Coeff} x^2 F(t, x)).$$

The same guess-and-check approach works for any finite set $S$. First we ask our beloved computer to use the functional equation to generate sufficiently many terms of the formal power series

$$F(0, t), \text{Coeff} x F(t, x), \ldots, \text{Coeff} x^r F(t, x),$$

where $r = -\min(S) - 1$. Then ask the computer to guess algebraic (or holonomic) descriptions for these, and call these conjectured representations $g_0(t), g_1(t), \ldots, g_r(t)$. Now define $G(t, x)$ by replacing $F(0, t)$ by $g_0(t)$, $\text{Coeff} x F(t, x)$ by $g_1(t)$ etc., and derive an algebraic (or holonomic) representation for $G(t, x)$. Now go backwards! Once you have an algebraic (or holonomic) description of $G(t, x)$, you also have algebraic (or holonomic) descriptions of $G(t, 0), \text{Coeff} x G(t, x)$, etc., and checking that (GeneralFunctionalEquation) with $F(t, x)$ replaced by $G(t, x)$ is a purely routine computation in the algebraic (or holonomic) ansatz!

Since we know that we can do it, why bother? Let’s just stick to the empirically-guessed algebraic (or holonomic) descriptions for $f(t)$, and declare it a theorem! Also, since $G(t, x)$ satisfies some (often, very complicated!) algebraic (and differential) equation with polynomial coefficients in $t$ and the parameter (‘catalytic’ variable) $x$, we know that $G(t, 1)$, the generating function for all good sequences (not necessarily totaling 0) is algebraic (and $D$-finite, and hence its sequence of coefficients, satisfies some linear recurrence with polynomial coefficients), it must be the same one that we guessed! So let’s declare it a theorem, and leave the proof to those (obtuse!) readers who care about mathematical certainty.
It turns out that, for this one-dimensional case, there is a ‘meta-theorem’, due to Phillippe Duchon ([D], see also [AZ]) that guarantees that everything in site is algebraic (and hence holonomic), so we know a priori, that there is some linear recurrence with polynomial coefficients satisfied by our enumerating sequence, and if we did not find it, it is just because we didn’t look far enough. Also once found, we are guaranteed that there exists a fully rigorous proof, but carrying out all the details would take much longer than the initial discovery of the ‘conjectured’ recurrences, hence it is a waste of time to actually do the full proof.

Two Dimensions and Beyond

For walks in two dimensions, staying in the first (non-negative) quadrant, with a prescribed set of steps, $S$, it is no longer guaranteed that the counting sequences are holonomic, as shown, in a seminal paper, by Marni Mishna and Andrew Rechnitzer [MR]. Nevertheless, there are many cases where it is holonomic, and some very smart people (See [BM] and its many references) are trying to understand why, and whenever possible, use human ingenuity (possibly assisted by computers), using such sophisticated methods as the kernel method (perfected in [BM]) to derive such equations.

Here too it is easy to set-up a functional equation in each case (but now we have two ‘catalytic’ variables, $x$ and $y$, in addition to $t$ (time), and the weight is $t^n x^i y^j$ where $n$, as before is the length of the walk, and $[i,j]$ is the end-point of the walk (that must be in the non-negative quadrant). If the set of steps is $S$, then the same reasoning as before shows that $F(t, x, y)$ satisfies the functional equation

$$F(t, x, y) = 1 + t \sum_{[s_1, s_2] \in S} x^{s_1} y^{s_2} \left( F(t, x) - \sum_{i=0}^{-s_1-1} \text{Coeff}_{x^i} F(t, x, y) - \sum_{j=0}^{-s_2-1} \text{Coeff}_{y^j} F(t, x, y) + \sum_{i=0}^{-s_1-1} \sum_{j=0}^{-s_2-1} \text{Coeff}_{x^i y^j} F(t, x, y) \right)$$

[Note that if $s_1 \geq 0$ or $s_2 \geq 0$, then the corresponding sums are empty.]

In each of the known cases, our naive guess-and-check approach works just as well! And with much less human effort! Not only that, we can do some cases where the kernel method fails. We now no longer have an a priori guarantee that the proof, in the holonomic ansatz, would succeed, but it is extremely unlikely, that once we succeeded in finding conjectured linear differential equations with polynomial coefficients satisfied by $F(t, 0, 0)$ and by $F(t, 1, 1)$, that $F(t, x, y)$ would not be a solution of another, more complicated, such linear differential equation with coefficients that are now polynomials in $x, y,$ and $t$. If such a differential equation, for $F(t, x, y)$, exists, then its proof is purely routine, but finding it (usually) takes too long, so why bother?

If you really want to play it safe, then try and find $D$-finite descriptions for, say $F(t, 2, 0)$ and $F(t, 0, 2)$ (since here we only have the variable $t$, this requires far less computational resources), and you may even venture, say $F(t, 2, 3)$. If all these special cases lead to $D$-finite descriptions, then you can be really sure.
Maple packages and Sample Input and Output Files

This article is accompanied by four Maple packages: \texttt{W1D}, for one-dimensional walks, \texttt{W1Dp}, for the probability analog, where the die may be loaded, \texttt{W2D}, for two-dimensional walks, and \texttt{W3D}, for three dimensions. They may all be gotten from the front of this article

http://www.math.rutgers.edu/~zeilberg/mamarim/mamarimhtml/gac.html

that contains links to many articles, generated by these packages, about walks for many sets of steps. For some simple cases (small sets in one dimensions), we present fully rigorous proofs, or at least sketches, but for the more complicated cases (for example for the set $S = \{-2, -1, 3\}$, quoted above), we did not bother, since we do not care.

Looking for Patterns

Francis Bacon, the pioneer of the Scientific method, believed that the scientist should study nature without any prejudice, find patterns, and then generalize, and formulate theories. But since the haystack is so big, people came to realize that there is always some pre-conceived ideas, and all observation is theory-laden, and one looks for specific types of patterns.

In the present case studies, we knew what kind of patterns to look for, namely the algebraic or holonomic ansatizes, but the specific patterns were way too complicated for the naked brain. So it usually takes much longer than a moment (see Wittgenstein’s quote in the motto), even for a powerful computer, to detect non-trivial patterns. It is our duty, as human coaches, to discover new ansatizes that would fit data not accounted by already known ansatizes, and then ask computers to search for patterns within them. One of the simplest ansatizes, that could have been used by 7-year-old Gauss, is the polynomial ansatz, and it is used often in IQ tests, where one is asked to continue a sequence, and a few such examples are given by Wittgenstein in his Philosophical Investigations.** But humans can only detect trivial and superficial patterns, so instead of looking for patterns themselves, they should teach the computer to look for them.

Conclusion: It’s time to Make Rigorous proofs Optional

With all due respect to counting walks, for us, it is but a case study to illustrate a class of problems where fully rigorous proofs can be safely abandoned. We believe that this would be the case, in at most fifty years, for the rest of mathematics. We will come to realize that fully rigorous proofs are only possible for relatively trivial statements, for example Fermat’s Last Theorem, the Poincaré conjecture, and the Four Color Theorem. But for really deep (and interesting!) mathematical knowledge, we would have to be content, if lucky, with semi-rigorous proofs, where we know that a proof exists, but it is too complicated for us, and even for our computers, to find it, and more

** Wittgenstein objected that one can continue a sequence arbitrarily and then make-up some ‘law’ that justifies it. But with the combination of Occam’s razor, and having specific ansatizes in mind, and only using, say, half of the known sequence for the ‘guessing’, and being able to confirm our guess with the remaining known values, resolves Wittgenstein’s objection.
often with fully non-rigorous (heuristic and empirical) proofs.

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