

Probability problems for Sigma - part 1

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Hello Sigma members!

After a long gap I have, at last, managed to find some time to revisit some mathematics and started reading an old book I have on probability. That book is “Probability - an Introduction” by Grimmet and Welsh (1985) from which I’ve investigated some of the problems/exercises from the first chapter which I hope you’ll find interesting. I’m starting this submission with a quick guide on recurrence relations (also based on a similar section from the same book) which seem to be useful in many situations when looking at probability problems. Hopefully, I’ll be looking at some extra problems for a subsequent newsletter.

1 A quick guide to solving linear recurrence relations

1.1 Introduction

This section only deals with techniques for solving recurrence relations (or “difference equations”) where we have a sequence of numbers x_0, x_1, \dots satisfying the relationship

$$a_0x_{n+k} + a_1x_{n+k-1} + \dots + a_kx_n = g(n) \quad \text{for } n = 0, 1, 2, \dots \quad (\text{R1})$$

where a_0, a_1, \dots, a_k are given real numbers and both $a_0 \neq 0$ and $a_k \neq 0$, and $g(n)$ is some polynomial function of n (or a constant).

To get an explicit solution, we also assume that we know the initial values x_0, x_1, \dots, x_{k-1} (although any k known values would suffice).

The first step is to find the roots of the *auxiliary equation* (sometimes called the ‘characteristic equation’) :

$$a_0\theta^k + a_1\theta^{k-1} + \dots + a_{k-1}\theta + a_k = 0 \quad (1)$$

If all of the roots (including complex roots) $\theta_1, \theta_2, \dots, \theta_k$ are distinct, then the general solution of (R1) (where $g(n) \equiv 0$ for now) is :

$$x_n = c_1\theta_1^n + c_2\theta_2^n + \dots + c_k\theta_k^n \quad \text{for } n = 0, 1, 2, \dots \quad (2)$$

and c_1, c_2, \dots, c_k are arbitrary constants which can be found by equating the first k values with the known initial values, and solving the simultaneous equations.

If any roots are the same (and occur m times), then the term for that root θ will instead be of the form $(s_1 + s_2n + \dots + s_mn^{m-1})\theta^n$

Finally, for $g(n) \not\equiv 0$ we firstly solve as above for $g(n) \equiv 0$ to obtain the *complementary solution* $x_n = \kappa_n$, then by any devious/clever means required, we find a *particular solution* for our particular $g(n)$, $x_n = \pi_n$

The general solution is then simply $x_n = \kappa_n + \pi_n$.

1.2 Example 1

Solve $x_{n+2} - 7x_{n+1} + 10x_n = 0$ where $x_0 = 2$ and $x_1 = 3$.

The auxiliary equation is $\theta^2 - 7\theta + 10 = 0 \implies \theta_1 = 2, \theta_2 = 5$ (the root order is not important).

The general solution is therefore of the form :

$$x_n = c_1 2^n + c_2 5^n \quad (3)$$

Setting $n = 0$ and $n = 1$ respectively, we have :

$$\begin{aligned} 2 &= c_1 + c_2 \\ 3 &= 2c_1 + 5c_2 \end{aligned}$$

which leads to $c_1 = \frac{7}{3}$ and $c_2 = \frac{-1}{3}$
The general solution is therefore :

$$x_n = \frac{7}{3} 2^n - \frac{1}{3} 5^n \quad (4)$$

1.3 Example 2

Solve $x_{n+3} - 5x_{n+2} + 8x_{n+1} - 4x_n = 0$ where $x_0 = 0$, $x_1 = 3$, $x_2 = 13$

The auxiliary equation is $\theta^3 - 5\theta^2 + 8\theta - 4 = 0 \implies \theta_1 = 2, \theta_2 = 2, \theta_3 = 2$ (again the root order is not important).

The general solution is therefore of the form :

$$x_n = c_1 1^n + (c_2 + c_3 n) 2^n \quad (5)$$

The boundary conditions can then be used to determine $c_1 = 1$, $c_2 = -1$ and $c_3 = 2$

1.4 Example 3

Solve $x_{n+2} - 5x_{n+1} + 6x_n = 4n + 2$ where $x_0 = 5$, $x_4 = -37$

As above, the complementary solution is of the form

$$x_n = c_1 2^n + c_2 3^n \quad (6)$$

By guesswork, trial and error, magic, instinct, or luck, we can find that $x_n = 2n + 4$ for $n = 0, 1, 2, \dots$ is a particular solution, so the general solution is therefore

$$x_n = c_1 2^n + c_2 3^n + 2n + 4 \quad (7)$$

The boundary conditions can then be used to determine $c_1 = 2$ and $c_2 = -1$.

2 Grimmet and Welsh Chapter 1 Problem 1

2.1 Question

A fair die is thrown n times. Show that the probability that there are an even number of sixes is

$$\mathbf{P}_n := \frac{1}{2} \left[1 + \left(\frac{2}{3} \right)^n \right]$$

2.2 Some general observations

Looking at specific values of n seems a good idea to get a feel for the behaviour of \mathbf{P}_n and to validate any formulae we obtain.

$$\mathbf{P}_0 = 0, \mathbf{P}_1 = \frac{5}{6}, \mathbf{P}_2 = \frac{1}{36} + \frac{25}{36} = \frac{26}{36}$$

For very large n , the number of sixes will be broadly spread around the value $\frac{n}{6}$ with broadly similar amounts of even and odd values, so we should also expect

$$\lim_{n \rightarrow \infty} \mathbf{P}_n = \frac{1}{2}$$

2.3 Answer 1

A direct approach!

We will use this standard result : If A_1, A_2, \dots are disjoint, i.e. $A_i \cap A_j = \emptyset$ for $i \neq j$:

$$\mathbf{P}\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mathbf{P}(A_i) \quad (\text{F1.1})$$

Firstly, consider when n is even (only to save messy-looking equations) :

Let $i\mathbf{P}_n$ be the probability of throwing i sixes from n throws.

Using (F1.1) :

$$\mathbf{P}_n = {}_0\mathbf{P}_n + {}_2\mathbf{P}_n + \dots + {}_n\mathbf{P}_n$$

But $i\mathbf{P}_n = \binom{n}{i} \left(\frac{1}{6}\right)^i \left(\frac{5}{6}\right)^{n-i}$, so

$$\mathbf{P}_n = \binom{n}{0} \left(\frac{1}{6}\right)^0 \left(\frac{5}{6}\right)^n + \binom{n}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{n-2} + \dots + \binom{n}{n} \left(\frac{1}{6}\right)^n \left(\frac{5}{6}\right)^0$$

To evaluate this, note that :

$$(1+x)^n + (1-x)^n = 2 \left(\binom{n}{0} x^0 + \binom{n}{2} x^2 + \dots + \binom{n}{n} x^n \right)$$

Setting $x = \frac{1}{5}$:

$$\frac{1}{2} \left[\left(\frac{6}{5}\right)^n + \left(\frac{4}{5}\right)^n \right] = \binom{n}{0} \left(\frac{1}{5}\right)^0 + \binom{n}{2} \left(\frac{1}{5}\right)^2 + \dots + \binom{n}{n} \left(\frac{1}{5}\right)^n$$

Multiplying by $\left(\frac{5}{6}\right)^n$ we get :

$$\mathbf{P}_n = \left(\frac{5}{6}\right)^n \times \left(\frac{1}{2}\right) \left(\left(\frac{6}{5}\right)^n + \left(\frac{4}{5}\right)^n \right) = \frac{1}{2} \left[1 + \left(\frac{2}{3}\right)^n \right]$$

For when n is odd, very similar logic applies. The revised equations corresponding to above are :

$$\begin{aligned} \mathbf{P}_n &= {}_0\mathbf{P}_n + {}_2\mathbf{P}_n + \dots + {}_{n-1}\mathbf{P}_n \\ \mathbf{P}_n &= \binom{n}{0} \left(\frac{1}{6}\right)^0 \left(\frac{5}{6}\right)^n + \binom{n}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{n-2} + \dots + \binom{n}{n-1} \left(\frac{1}{6}\right)^{n-1} \left(\frac{5}{6}\right)^1 \\ (1+x)^n + (1-x)^n &= 2 \left(\binom{n}{0} x^0 + \binom{n}{2} x^2 + \dots + \binom{n}{n-1} x^{n-1} \right) \end{aligned}$$

and the same result is obtained.

2.4 Answer 2

Using a recurrence relation.

If we throw a die $n + 1$ times then to obtain an even number of sixes, either the first throw is not a six (with probability $5/6$) and we need an even number of sixes from the remaining n throws, or the first throw is a six (with probability $1/6$) and we need an odd number of sixes from the remaining n throws.

So, conditioning on the first throw :

$$\mathbf{P}_{n+1} = \frac{5}{6}\mathbf{P}_n + \frac{1}{6}(1 - \mathbf{P}_n) \Rightarrow \mathbf{P}_{n+1} = \frac{1}{6} + \frac{2}{3}\mathbf{P}_n$$

Using “standard” methods, the auxiliary equation is $\theta - \frac{2}{3} = 0$, yielding the complementary solution :

$$A \left(\frac{2}{3} \right)^n$$

and a particular solution $\mathbf{P}_n = \frac{1}{2}$

So, the general solution is $\mathbf{P}_n = \frac{1}{2} + A \left(\frac{2}{3} \right)^n$, but as we know $P_0 = 1$, then $A = \frac{1}{2}$.

Therefore, as before :

$$\mathbf{P}_n = \frac{1}{2} \left[1 + \left(\frac{2}{3} \right)^n \right]$$

2.5 Answer 3

Given that we are provided with the answer, we can use induction on n .

This is easiest but least pleasing (as induction often is) as we don’t really gain any insight into why it is true. . . .

The statement is clearly true for $n = 0$ as $\mathbf{P}_0 = 0$.

Assume it is true for $n = k$. Therefore :

$$\mathbf{P}_k = \frac{1}{2} \left[1 + \left(\frac{2}{3} \right)^k \right]$$

But, as in Answer 2,

$$\mathbf{P}_{k+1} = \frac{1}{6} + \frac{2}{3}\mathbf{P}_k = \frac{1}{6} + \frac{2}{3} \times \frac{1}{2} \times \left[1 + \left(\frac{2}{3} \right)^k \right] = \frac{1}{2} \left[1 + \left(\frac{2}{3} \right)^{k+1} \right]$$

so it is also true for $n = k + 1$ and so true for all $n \geq 0$.

2.6 Final comments

As expected, the formula does indeed tend to $\frac{1}{2}$ as $n \rightarrow \infty$. Also, an obvious consequence is that the probability that there are an odd number of sixes in n throws is

$$\mathbf{P}_n = \frac{1}{2} \left[1 - \left(\frac{2}{3} \right)^n \right]$$

3 Grimmet and Welsh Chapter 1 Exercise 27

3.1 Question

A biased coin shows heads with probability $p = 1 - q$ whenever it is tossed.

Let u_n be the probability that, in n tosses, no pair of heads occur successively. Show that for $n \geq 1$

$$u_{n+2} = qu_{n+1} + pqu_n$$

and find u_n by the usual method if $p = \frac{2}{3}$

3.2 Answer 1

We will be using the “Partition Theorem” :

If $B_i \cap B_j = \emptyset$ and $\cup_i(B_i) = \Omega$ and $\mathbf{P}(B_i) > 0 \quad \forall i :$

$$\mathbf{P}(A) = \sum_i \mathbf{P}(A | B_i) \mathbf{P}(B_i) \quad \text{for all } A \quad (\text{F1.9})$$

Let A_n represent the event that, in n tosses, no pair of heads occur successively. Let B_i be the event that the first $(i - 1)$ tosses are heads (“H”) and the i th toss is tails (“T”) (and $i \geq 1$)

So B_1 represents the events $T \dots$,

B_2 represents the events $HT \dots$,

B_3 represents the events $HHT \dots$ etc.

Note that if Ω represents all possible events

$$\bigcup_i B_i = \Omega \text{ and } B_i \cap B_j = \emptyset \text{ for all } i \neq j$$

So, we can use the Partition Theorem (F1.9) as follows.

$$u_{n+2} = \mathbf{P}(A_{n+2}) = \mathbf{P}(A_{n+2} | B_1) \mathbf{P}(B_1) + \mathbf{P}(A_{n+2} | B_2) \mathbf{P}(B_2) + \dots$$

But, $\mathbf{P}(A_{n+2} | B_1)$ looks at the event given a beginning “T”, which ‘resets’ the starting state to having $n + 1$ tosses in which to get no pair of heads (since the tail does not affect the situation for subsequent throws). This also applies to $\mathbf{P}(A_{n+2} | B_2)$.

$\mathbf{P}(A_{n+2} | B_3)$, $\mathbf{P}(A_{n+2} | B_4) \dots$ are all 0 since the outcome of no pair of heads has failed for all of them. So,

$$= u_{n+1}q + u_n pq + 0 + 0 + \dots$$

For the second part, we substitute $p = \frac{2}{3}$ to obtain

$$u_{n+2} = \frac{1}{3}u_{n+1} + \frac{2}{9}u_n \implies 9u_{n+2} - 3u_{n+1} - 2u_n = 0$$

and we also note that $u_1 = 1$, and $u_2 = \frac{5}{9}$

Using “standard” methods, the auxiliary equation is $9\theta^2 - 3\theta - 2 = 0 \rightarrow (3\theta + 1)(3\theta - 2) = 0$, yielding the general solution :

$$u_n = A \left(\frac{-1}{3} \right)^n + B \left(\frac{2}{3} \right)^n \text{ for some } A \text{ and } B.$$

$$u_1 = 1 \implies -\frac{1}{3}A + \frac{2}{3}B = 1 \rightarrow 2B - A = 3$$

$$u_2 = \frac{5}{9} \implies A + 4B = 5$$

which leads to

$$B = \frac{4}{3} \text{ and } A = -\frac{1}{3}$$

$$\therefore u_n = \frac{4}{3} \left(\frac{2}{3} \right)^n - \frac{1}{3} \left(\frac{-1}{3} \right)^n \text{ for } n \geq 1 \quad (8)$$

There are other approaches considered later.

3.3 Comments

Below is a table of probabilities for the first few values of n . As it is easy to make mistakes deriving formulae, it’s a good idea to verify results by using a separate mechanism - in this case, a simple Java program which ‘makes’, for each n , n tosses of a coin millions of times and counts the proportion which don’t contain 2 consecutive heads.

As you can see, the results suggest that the answers above are (probably) correct!

Also note the successive ratio of probabilities for increasing values of n which, as expected, converge to the dominant power $2/3$ in (8) above.

3.4 Alternative solution

We can also get the same result by letting v_n be the probability that, in n tosses, a pair of heads does occur successively. Defining A_n^* as the event that, in n tosses, a pair of heads does occur successively, and B_i exactly as before,

$$\begin{aligned} v_{n+2} &= \mathbf{P}(A_{n+2}^*) = \mathbf{P}(A_{n+2}^* \mid B_1)\mathbf{P}(B_1) + \mathbf{P}(A_{n+2}^* \mid B_2)\mathbf{P}(B_2) + \dots \\ &= v_{n+1}q + v_npq + 1 \cdot pq + 1 \cdot p^2q + 1 \cdot p^3q + \dots \\ &= v_{n+1}q + v_npq + \frac{p^2q}{1-p} = v_{n+1}q + v_npq + p^2 \end{aligned}$$

from which we can either substitute $u_n = 1 - v_n$ to get the previous relation, or proceed to solve it for a particular value of p and substitute later...

Solving for $p = \frac{2}{3}$ we obtain

$$v_n = 1 - \frac{4}{3} \left(\frac{2}{3}\right)^n + \frac{1}{3} \left(\frac{-1}{3}\right)^n \equiv 1 - u_n \quad (9)$$

4 Grimmet and Welsh Chapter 1 Problem 7

4.1 Question

Two people toss a fair coin n times. Show that the probability that they throw an equal number of heads is

$$\binom{2n}{n} \left(\frac{1}{2}\right)^{2n}$$

4.2 Answer 1

The hints section of the book suggests this.

If X and Y are the numbers of heads tossed by each person :

$$\mathbf{P}(X = Y) = \sum_k \mathbf{P}(X = k)\mathbf{P}(Y = k) = \sum_k \mathbf{P}(X = k)\mathbf{P}(Y = n - k)$$

The last statement works because the coin is fair, so this is essentially saying that k heads has the same probability as k tails, which is the same event as $n - k$ heads. But,

$$\sum_k \mathbf{P}(X = k)\mathbf{P}(Y = n - k) = \mathbf{P}(X + Y = n)$$

which is the probability of, regardless of how they are distributed amongst each person, n heads result from $2n$ tosses, which is just $\binom{2n}{n} \left(\frac{1}{2}\right)^{2n}$

n	$\frac{4}{3} \left(\frac{2}{3}\right)^n - \frac{1}{3} \left(\frac{-1}{3}\right)^n$	Ratio	Java program
1	1		1
2	0.5555555556	0.555556	0.55553827
3	0.4074074074	0.733333	0.407404815
4	0.2592592593	0.636364	0.25929586
5	0.1769547325	0.68254	0.176948485
6	0.1165980796	0.658915	0.116604615
7	0.0781893004	0.670588	0.078175775
8	0.0519737845	0.664717	0.05198466
9	0.0346999949	0.667644	0.034695565
10	0.0231163949	0.666179	0.023142995
11	0.0154165749	0.666911	0.01540314
12	0.0102758349	0.666545	0.01027086

Here is a more direct (and longer!) approach.

Now,

If we multiply these two equations, but only look at the coefficients of x^n we obtain :

and the result follows immediately. \square

The above equations have an interesting 'visual' appeal when we look at Pascal's triangle.

This works for any such inverted triangle.

Using Stirling's formula :

This is a very good approximation for the actual values, even for small n .

4.6 What about an unfair coin?

We cannot use the sneaky trick from Answer 1, so we can perhaps attempt a direct approach again. . . We now use a biased coin which shows heads with probability $p = 1 - q$ whenever it is tossed.

Now,

$$\mathbf{P}(X = Y) = \sum_k \mathbf{P}(X = k) \mathbf{P}(Y = k) = \sum_k \left(\binom{n}{k} \right)^2 p^{2k} q^{2n-2k}$$

If we try a similar trick to evaluate this, we would start with :

$$(1 + p^2 x)^n = \binom{n}{0} x^0 + \binom{n}{1} p^2 x^1 + \binom{n}{2} p^2 x^2 + \cdots + \binom{n}{n-1} p^{2n-2} x^{n-1} + \binom{n}{n} p^{2n} x^n, \text{ and}$$

$$(q^2 x + 1)^n = \binom{n}{0} q^{2n} x^n + \binom{n}{1} q^{2n-2} x^{n-1} + \binom{n}{2} q^{2n-4} x^{n-2} + \cdots + \binom{n}{n-1} q^2 x^1 + \binom{n}{n} x^0$$

and we could try to determine the coefficient of x^n in $(1 + p^2 x)^n (1 + q^2 x)^n$

Unfortunately, there appears to be no naturally simple answer to this (except for the sum we are trying to determine in the first place).

I suspect there is no closed formula for the probability for the unfair coin, but I'm very happy to be corrected by any readers!