

MATHEMATICS AND STATISTICS RESEARCH COMPETITION

QUESTION 8

JIAMU LI & FRANK TANG & EDWARD WANG

SCOTCH COLLEGE

JULY 31, 2022

CONTENTS

Problem 1	1
Problem 2	4
Problem 3	5
Generalisation	7
An interesting special case	10
Limiting difference between terms	11
Code Implementation	13
Code Implementation for Generalisation	15
References	18

A particle generator is emitting two types of particles (called X and Y) into a long tube. The particles will line up in order after entering the tube. Initially, the tube is empty. At each shot, either an X- or Y-particle is randomly emitted into the tube with equal probability. Different shots are assumed to be independent from each other. Suppose that n shots have been emitted.

PROBLEM 1

- What is the probability that no two X-particles are next to each other?

THEOREM 1. *The probability that no two X-particles are next to each other after n shots is given by*

$$\frac{F_{n+2}}{2^n},$$

where F_n is the n th Fibonacci number.

Proof. The probability we require can be calculated by dividing the total number of ways to arrange the contents of the tube such that there are no consecutive X-particles, by the total number of arrangements of the particles. That is to say:

$$\Pr(\text{No consecutive X-particles}) = \frac{\# \text{Arrangements w/o consecutive X-particles}}{\# \text{Total arrangements}}$$

CLAIM 1.1. *The number of arrangements with no consecutive X-particles is*

$$\sum_{k=0}^n \binom{n-k+1}{k}. \quad (1)$$

Proof. Consider a tube with n particles in it. Let the number of X-particles be equal to k , so that the number of Y-particles is $n - k$. Consider the tube without the X-particles, consisting solely of Y-particles in a line:

$$\underbrace{\text{YY} \dots \text{YY}}_{n-k}$$

Now consider the ‘gaps’ between these Y-particles, indicated by a bar:

$$|Y|Y|\dots|Y|Y|$$

Notice that there are exactly $n - k + 1$ ‘gaps’. Clearly, if we were to only place (single) X-particles in the gaps, then there would never be any consecutive X-particles. This can be done in a total of

$$\binom{n-k+1}{k}$$

ways. However, we must consider this for any number of X-particles k , so we arrive at the sum

$$\# \text{Arrangements with no consecutive X-particles} = \sum_{k=0}^n \binom{n-k+1}{k}. \quad \square$$

CLAIM 1.2.

$$\sum_{k=0}^n \binom{n-k+1}{k} = F_{n+2},$$

where F_n is the n th Fibonacci number.

Proof. Figure 1 showcases a way to obtain the identity from Pascal's triangle. We will present an algebraic proof below.

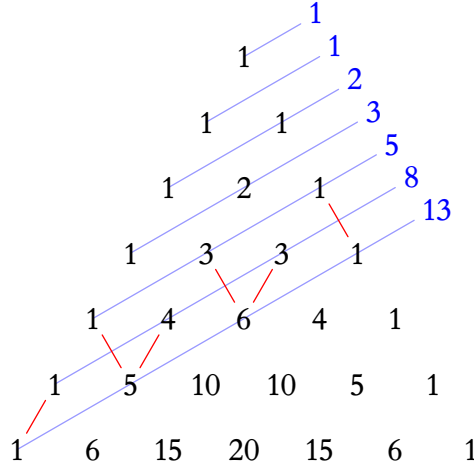


Figure 1: Each number in the row is the sum of the numbers in the previous two rows.

Recall that the Fibonacci numbers are defined as follows:

$$\begin{aligned} F_0 &= 0, \\ F_1 &= 1, \\ F_n &= F_{n-1} + F_{n-2}. \end{aligned} \quad (n > 1)$$

We will define a function $f(n) = \sum_{k=0}^n \binom{n-k+1}{k}$. It is sufficient to prove that $f(1) = F_3 = 2$, $f(2) = F_4 = 3$, and that $f(n) = f(n-1) + f(n-2)$, which would then imply the result by definition of the Fibonacci numbers.

It is obvious that $f(1) = \binom{2}{0} + \binom{1}{1} = 2$, which is equal to F_3 . Next, $f(2) = \binom{3}{0} + \binom{2}{1} + \binom{1}{2} = 3$. Notice that we define $\binom{n}{k} = 0$ when $n < k$, as it is impossible to choose k things from a set with less than k elements.

We proceed to prove that $f(n) = f(n-1) + f(n-2)$, where $n > 2$. Using the fact that $\binom{n}{0} = 1$, we rewrite $f(n)$ using Pascal's identity and linearity as

$$f(n) = 1 + \sum_{k=1}^n \binom{n-k+1}{k} = 1 + \sum_{k=1}^n \binom{n-k}{k} + \sum_{k=1}^n \binom{n-k}{k-1}.$$

Next, we simplify, getting

$$\begin{aligned}
 f(n) &= \sum_{k=0}^n \binom{n-k}{k} + \sum_{k=1}^n \binom{n-k}{k-1} \\
 &= \sum_{k=0}^{n-1} \binom{n-k}{k} + \binom{n-n}{n} + \sum_{k=1}^n \binom{n-k}{k-1} \\
 &= \sum_{k=0}^{n-1} \binom{n-k}{k} + 0 + \sum_{k=0}^{n-1} \binom{n-k-1}{k} \\
 &= \sum_{k=0}^{n-1} \binom{n-k}{k} + \sum_{k=0}^{n-2} \binom{n-k-1}{k} + \binom{n-(n-1)-1}{n-1} \\
 &= f(n-1) + f(n-2) + 0.
 \end{aligned}$$

Hence $f(n)$ must be equivalent to F_{n+2} . \square

In the spirit of Arthur Benjamin and Jennifer Quinn[1], we also provide a purely combinatorial proof for Claim 1.2.

Proof. Let the number of arrangements without consecutive X-particles after n shots be R_n . Now consider the two choices of the first particle: either Y or X. If the first particle was a Y-particle, then the number of arrangements without consecutive X-particles is simply R_{n-1} . However, if the first particle was an X-particle, then the next particle must be a Y-particle, and therefore the number of arrangements is R_{n-2} . Hence $R_n = R_{n-1} + R_{n-2}$, and since $R_1 = 2$ and $R_2 = 3$, we have $R_n = F_{n+2}$. We then equate this with the formula obtained in Claim 1.1 to conclude the proof. \square

CLAIM 1.3. *The number of total arrangements of a tube with n particles is*

$$2^n.$$

Proof. Each particle in the tube can be either an X-particle or a Y-particle, meaning there are 2 choices for each of the n particles. Hence, there are a total of 2^n arrangements. \square

Hence by dividing the number of arrangements where there are no consecutive X-particles by the total number of arrangements, we arrive at the formula

$$\frac{F_{n+2}}{2^n}$$

which gives the desired probability. \square

PROBLEM 2

Now, if two X-particles are next to each other, they will immediately collide into one single X-particle.

- Compute the average number of particles for $n = 2, 3, 4$.
- Can you find the pattern and establish an explicit formula for general n ?

We found it most straightforward to proceed directly to finding a general formula. However, it should be noted that it is relatively simple to compute the averages for small n by considering every possible tube after n shots.

THEOREM 2. *The average number of particles after n shots is*

$$\frac{3}{4}n + \frac{1}{4}.$$

Proof. Let the average number of particles after n shots be T_n . Obviously, $T_1 = 1$. Next, consider the chance that after firing a shot, the number of particles *doesn't* increase. This occurs only in the event that the last particle in the tube is an X-particle, and when the particle emitted is also an X-particle. Since both events have a probability of $1/2$, the probability that both occur is simply $1/4$.

$$\begin{aligned} X &\rightarrow X : X \\ Y &\rightarrow X : YX \\ X &\rightarrow Y : XY \\ Y &\rightarrow Y : YY \end{aligned}$$

Hence, the probability that the number of particles *does* increase after firing a shot is $1 - 1/4 = 3/4$. Thus the expected number of particles increases by $3/4$ after each shot, giving us the recursion

$$T_{n+1} = T_n + \frac{3}{4}.$$

Since $T_1 = 1$, we arrive at the formula

$$T_n = \frac{3}{4}n + \frac{1}{4}. \quad \square$$

Using this, we can easily compute the average number of particles when $n = 2, 3, 4$:

n	T_n
1	1
2	1.75
3	2.5
4	3.25

PROBLEM 3

Suppose that at each shot, an X-particle is emitted with probability $p \in (0, 1)$.

- Under the same assumption as above, when n is very large, do you think the proportion of X-particles in the tube will eventually stabilise at a certain number? Why/why not?
- If so, can you compute this number explicitly?

THEOREM 3. *Suppose the probability of emitting an X-particle is $p \in (0, 1)$. The limiting ratio of X-particles in the tube, as the number of shots approaches infinity, is*

$$\frac{p}{1+p}.$$

Proof. We use a similar argument to that which is featured in Theorem 2 to arrive at a formula for the expected number of particles after n shots when the probability is p .

CLAIM 3.1. *The average number of particles after n shots when the probability of emitting an X-particle is $p \in (0, 1)$ is*

$$(1 - p^2)n + p^2.$$

Proof. Again, let the average number of particles after n shots be T_n . The number of particles *doesn't* increase when the last particle is an X-particle and the particle emitted is also an X-particle, which has a probability of p^2 of occurring. Hence, the probability that the number of particles *does* increase is $1 - p^2$, meaning that the expected number of particles increases by $1 - p^2$ after each shot, and thus we obtain

$$T_{n+1} = T_n + (1 - p^2).$$

Since $T_1 = 1$, the formula for T_n is

$$T_n = (1 - p^2)n + p^2. \quad \square$$

We can now find the expected number of X-particles in the tube. Since the expected number of Y-particles in the tube is simply $(1 - p)n$, as the probability of emitting a Y-particle is $(1 - p)$ at each shot, the expected number of X-particles is

$$\begin{aligned} \text{\#X-particles} &= \text{\#Total particles} - \text{\#Y-particles} \\ &= T_n - (1 - p)n \\ &= (1 - p^2)n + p^2 - (1 - p)n. \end{aligned}$$

We then divide this by the total number of particles to obtain the desired proportion and then take the limit as $n \rightarrow \infty$, to obtain

$$\lim_{n \rightarrow \infty} \frac{(1 - p^2)n + p^2 - (1 - p)n}{(1 - p^2)n + p^2} = \lim_{n \rightarrow \infty} \frac{(1 - p^2) + \frac{p^2}{n} - (1 - p)}{(1 - p^2) + \frac{p^2}{n}}.$$

By the algebraic limit theorem, we obtain

$$\begin{aligned}\frac{(1-p^2)+0-(1-p)}{(1-p^2)+0} &= \frac{p-p^2}{1-p^2} \\ &= \frac{p(1-p)}{(1+p)(1-p)} \\ &= \frac{p}{1+p}\end{aligned}\quad \square$$

GENERALISATION

We were able to produce a generalisation of Problem 2:

- Suppose that if m X-particles are next to each other, they collapse into n X-particles.
- The probability of firing an X-particle is $p \in [0, 1]$.
- Assume that k particles have been emitted.
- Find the average number of particles.
- It is assumed that $m > n \geq 1$, otherwise the number of particles would simply blow up infinitely.

THEOREM 4. *The average number of particles after k shots, when m X-particles collapse into n , is k when $k < m$, and is*

$$m - 1 + \sum_{b=m-1}^{k-1} \left(1 - p \left(\sum_{a=0}^{\lfloor \frac{b-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1} + \varepsilon \right) (m-n) \right) \quad (2)$$

$$\text{when } k \geq m \text{ and } \varepsilon = \begin{cases} p^b, & b - m + 1 \equiv 0 \pmod{m-n} \\ 0, & b - m + 1 \not\equiv 0 \pmod{m-n}. \end{cases}$$

Proof. Let the average number of particles after k shots be denoted by T_k . It is obvious that $T_k = k$ when $k < m$, as it is impossible for any particles to have collapsed. Thus we must proceed to find a formula for T_k when $k \geq m$. We can achieve this by establishing a recurrence for T_k in order to find the difference between each term, and from there we may simply sum the difference to obtain the required formula.

We know that T_k is recursive since there are two events that can occur after each shot:

- Number of particles increases by 1
- Number of particles increases by $n - m + 1$

This is because if an X-particle were added to $m - 1$ consecutive X-particles, they would collapse into n X-particles. Hence, the number of particles ‘increases’ by $n - m + 1$, although this number is always negative or 0 (as $m > n$). Thus it remains to determine the probability that there are $m - 1$ consecutive X-particles at the end of the tube. Letting this probability be ϑ , the probability of the particles collapsing on the next shot is simply $p\vartheta$. As such, the expected number of particles increases by 1 with a probability of $1 - p\vartheta$ and increases by $n - m + 1$ with a probability of $p\vartheta$. Hence, the recurrence is:

$$\begin{aligned} T_{k+1} &= T_k + 1 - p\vartheta + p\vartheta(n - m + 1) \\ &= T_k + 1 - p\vartheta(m - n). \end{aligned} \quad (3)$$

The remaining step is to calculate ϑ .

CLAIM 4.1.

$$\vartheta = \sum_{a=0}^{\lfloor \frac{k-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1} + \varepsilon,$$

$$\text{where } \varepsilon = \begin{cases} p^k, & k-m+1 \equiv 0 \pmod{m-n} \\ 0, & k-m+1 \not\equiv 0 \pmod{m-n}. \end{cases}$$

Proof. Recall that ϑ is the probability that there are $m-1$ consecutive X-particles at the end of the tube. In order to ‘count’ the number of consecutive X-particles, there must be some end to the string of X-particles. This ‘end’ can occur in two ways: either there is a Y-particle before the last X-particle; or there are no Y-particles at all, and the tube comprises completely of X-particles.

$$\underbrace{\text{X} \dots \text{X}}_{m-1} \text{Y} \tag{A}$$

$$\underbrace{\text{X} \dots \text{X}}_{m-1} \tag{B}$$

Let us first consider the case with a Y-particle at the end (configuration A). The probability of such a configuration occurring is simply $(1-p)p^{m-1}$. However, this probability does not account for possible previous collapses. For example, the following sequence of particles may have been emitted which would lead to the same configuration:

$$\underbrace{\text{X} \dots \text{X}}_{m+(m-n)-1} \text{Y} \longrightarrow \underbrace{\text{X} \dots \text{X}}_{n+(m-n)-1} \text{Y} = \underbrace{\text{X} \dots \text{X}}_{m-1} \text{Y}$$

It is clear that adding any multiple of $m-n$ X-particles to $m-1$ consecutive X-particles results in the exact same configuration. However, since only k particles have been fired, $a(m-n) + m - 1 + 1 \leq k$ for some positive integer a . We solve this inequality for a , getting

$$\begin{aligned} a(m-n) + m &\leq k \\ a(m-n) &\leq k - m \\ a &\leq \frac{k-m}{m-n}. \end{aligned}$$

Thus the maximum value of a is $\left\lfloor \frac{k-m}{m-n} \right\rfloor$, so the probability of configuration A is

$$\sum_{a=0}^{\lfloor \frac{k-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1}.$$

We now consider configuration B, which consists of X-particles only. Again, the configuration can be achieved with $a(m-n) + m - 1$ consecutive X-particles, for some positive integer a . However, this time there cannot be any other particles other than these X-particles, meaning that $k = a(m-n) + m - 1$. It now becomes obvious that $k - m + 1$ must be an integer multiple of $m - n$, at which point there is a single

sequence of particles which results in $m - 1$ consecutive X-particles. When $k - m + 1$ is not an integer multiple of $m - n$, there is no possible way for configuration B to exist. Thus the probability of configuration B occurring can be expressed by ε , where

$$\varepsilon = \begin{cases} p^k, & k - m + 1 \equiv 0 \pmod{m - n} \\ 0, & k - m + 1 \not\equiv 0 \pmod{m - n}. \end{cases}$$

We then add the probabilities of configurations A and B to get the overall probability of the tube ending in $m - 1$ consecutive X-particles, which is

$$\vartheta = \sum_{a=0}^{\lfloor \frac{k-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1} + \varepsilon,$$

as required. \square

Since we have previously determined a recursive formula for T_k in terms of ϑ and ε in Equation 3, we can simply sum the difference between each term to obtain a closed expression for the value of any term. We have

$$T_{k+1} = T_k + 1 - p\vartheta(m - n),$$

meaning the difference between terms is $1 - p\vartheta(m - n)$. We then add $m - 1$ to the sum of this difference from $m - 1$ to $k - 1$ (since the formula is only required for $k \geq m$). This yields

$$T_k = m - 1 + \sum_{b=m-1}^{k-1} \left(1 - p \left(\sum_{a=0}^{\lfloor \frac{b-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1} + \varepsilon \right) (m - n) \right). \quad \square$$

It should be noted that ϑ is the sum of a geometric sequence, although for brevity, the sum is left unexpanded. The expanded sum is

$$\vartheta = \frac{(p-1)p^{m+n-1} \left(p^{(m-n)(\lfloor \frac{k-m}{m-n} \rfloor + 1)} - 1 \right)}{p^n - p^m}.$$

In addition, there also exists a closed form of ε , though it is somewhat distasteful. If we define a summation as 0 when its upper bound is less than its lower bound, then we obtain

$$\varepsilon = \sum_{a=0}^{(m-k-1) + \lfloor \frac{k-m+1}{m-n} \rfloor \cdot (m-n)} p^k.$$

The upper bound is simply the negative remainder of $k - m + 1$ when divided by $m - n$, which is 0 if and only if $k - m + 1$ is an integer multiple of $m - n$, and negative everywhere else. The lower bound, a is 0 simply to make the sum 0 whenever the upper bound is negative, and is not used elsewhere.

AN INTERESTING SPECIAL CASE

The general formula for the average length in Equation 2 is rather unwieldy. However, observe that many of the terms depend on $m - n$. If we set $m - n = 1 \implies n = m - 1$, then the formula can be reduced dramatically:

$$\begin{aligned} T_k &= m - 1 + \sum_{b=m-1}^{k-1} \left(1 - p \left(\sum_{a=0}^{\lfloor \frac{b-m}{1} \rfloor} (1-p)p^{a(1)+m-1} + \varepsilon \right) (m-n) \right) \\ &= m - 1 + \sum_{b=m-1}^{k-1} \left(1 - p \left(\sum_{a=0}^{b-m} (1-p)p^{a+m-1} + p^b \right) \right) \end{aligned} \quad (4)$$

Notice that even $\varepsilon = p^b$ becomes constant, as $z \equiv 0 \pmod{1}$ for any integer z . Let us now evaluate ϑ , which is the sum of a geometric sequence:

$$\begin{aligned} \vartheta &= \sum_{a=0}^{k-m} (1-p)p^{a+m-1} + p^k \\ &= \sum_{a=0}^{k-m} (1-p)(p^{m-1})p^a + p^k \\ &= (1-p)(p^{m-1}) \left(\frac{1-p^{k-m+1}}{1-p} \right) + p^k \\ &= p^{m-1}(1-p^{k-m+1}) + p^k \\ &= p^{m-1} - p^k + p^k \\ &= p^{m-1} \end{aligned}$$

This is dependent only on m , meaning ϑ is constant. Hence the difference between the terms of T_k is constant when $n = m - 1$, and so T_k would be linear. Let us now compute this difference more precisely. Using Equation 4, we see that the difference is

$$1 - p\vartheta = 1 - p^m.$$

The formula for T_k is then simply

$$\begin{aligned} T_k &= m - 1 + \sum_{b=m-1}^{k-1} (1 - p^m) \\ &= m - 1 + (k - m + 1)(1 - p^m). \end{aligned} \quad (5)$$

Using Equation 5, we can easily derive the formula for T_k when $m = 2$, $n = 1$ and $p = \frac{1}{2}$, which is equivalent to Problem 2:

$$\begin{aligned} T_k &= 2 - 1 + (k - 2 + 1) \left(1 - \frac{1}{4} \right) \\ &= 1 + \frac{3}{4}(k - 1) \\ &= \frac{3}{4}k + \frac{1}{4}. \end{aligned}$$

LIMITING DIFFERENCE BETWEEN TERMS

We now consider the difference between the terms, as the number of shots, k , tends towards infinity. Recall that the difference is

$$T_{k+1} - T_k = 1 - p\vartheta(m-n), \quad (6)$$

where

$$\vartheta = \sum_{a=0}^{\lfloor \frac{k-m}{m-n} \rfloor} (1-p)p^{a(m-n)+m-1} + \varepsilon$$

and

$$\varepsilon = \begin{cases} p^k, & k-m+1 \equiv 0 \pmod{m-n} \\ 0, & k-m+1 \not\equiv 0 \pmod{m-n}. \end{cases}$$

As k increases, ε tends towards 0 since its only non-zero value is p^k and $p < 1$. In addition, $\lfloor \frac{k-m}{m-n} \rfloor$ tends towards infinity. We hence have the limit as k approaches infinity of

$$\vartheta = \sum_{a=0}^{\infty} (1-p)p^{a(m-n)+m-1}.$$

Evaluating this geometric series yields

$$\begin{aligned} \vartheta &= \sum_{a=0}^{\infty} (1-p)(p^{m-1})(p^{m-n})^a \\ &= \frac{(1-p)p^{m-1}}{1-p^{m-n}}. \end{aligned}$$

Plugging this into Equation 6 gives us

$$\begin{aligned} \lim_{k \rightarrow \infty} T_{k+1} - T_k &= 1 - p \left(\frac{(1-p)p^{m-1}}{1-p^{m-n}} \right) (m-n) \\ &= 1 - \left(\frac{p^m - p^{m+1}}{1-p^{m-n}} \right) (m-n). \end{aligned} \quad (7)$$

Therefore the difference converges to a constant as the number of shots increases, implying that T_k becomes more and more linear as the number of terms increases.

For example, let us calculate this limiting difference for $m = 4$, $n = 1$, $p = \frac{1}{2}$. Plugging in these values, we obtain $\frac{25}{28}$. By plotting the difference between terms, we see that the difference does indeed converge to this value (Figure 2).

We can also plot the average number of particles as the number of shots increases, whilst varying p . Obviously, when $p = 0$, the line is simply linear. However, things get more interesting as p gets very close to 1. A plot is shown in Figure 3, in which the average number of particles can be seen oscillating when $p = 1$. This oscillation is caused by the constant emission of X-particles, which causes it to collapse every $m-n$ shots. In addition, notice that $0 \leq p \leq 0.5$ produce very similar values.

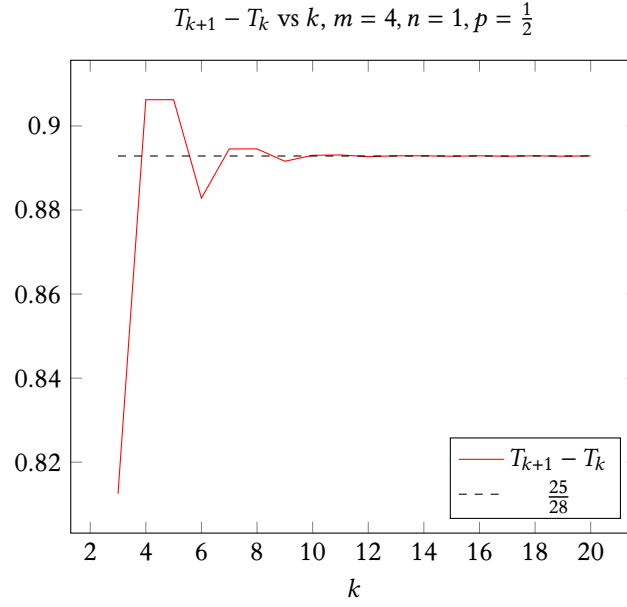


Figure 2

Average number of particles vs k , colour corresponds to p , $m = 5, n = 3$

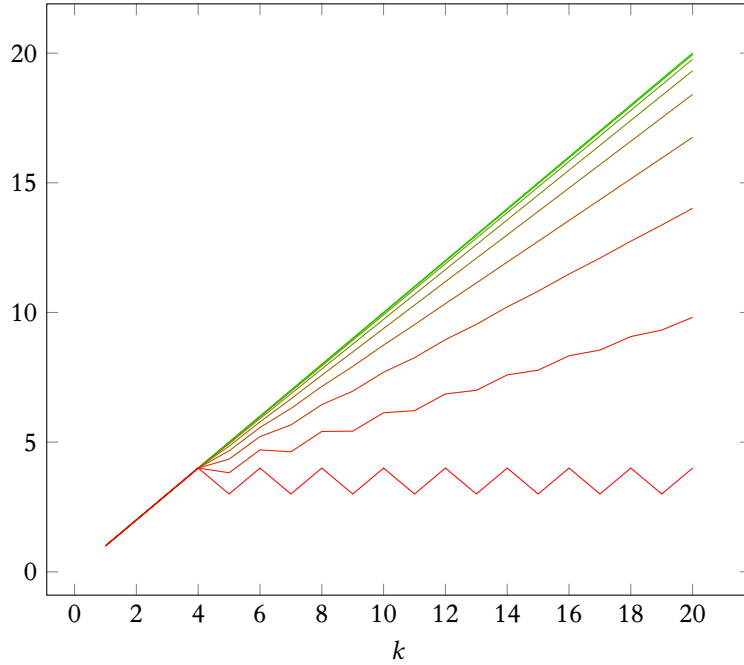


Figure 3: Pure green corresponds to $p = 0$, pure red corresponds to $p = 1$, increments of 0.1

CODE IMPLEMENTATION

We provide C++ source code for programs used to check against theoretically determined values.

```

1  #include <future>
2  #include <iostream>
3  #include <random>
4  #include <string>
5  #include <thread>
6  #include <vector>
7
8  using namespace std;
9
10 random_device rd;
11 mt19937 rng(rd());
12
13 const char particles[] = "XY";
14
15 unsigned int num_threads = thread::hardware_concurrency();
16
17 string gen_tube(int length, double p) {
18     discrete_distribution<int> pick{p, 1 - p};
19     string tube;
20     for (int i = 0; i < length; i++) {
21         tube += particles[pick(rng)];
22     }
23     return tube;
24 }
25
26 string annihilate(string tube) {
27     string output = "";
28     for (int i = 0; i < tube.length() - 1; i++) {
29         if (tube[i] != tube[i + 1] && tube[i] == 'X')
30             output += tube[i];
31         else if (tube[i] != 'X')
32             output += tube[i];
33     }
34     output.push_back(tube.back());
35     return output;
36 }
37
38 bool check_consec_x(string tube) {
39     for (int i = 0; i < tube.length() - 1; i++) {
40         if (tube[i] == tube[i + 1] && tube[i] == 'X')
41             return 0;
42     }
43     return 1;
44 }
45
46 double prob_no_two_consec_after_n(int n, int runs) {
47     int count = 0;
48     for (int i = 0; i < runs; i++) {
49         if (check_consec_x(gen_tube(n, 0.5)))
50             count++;
51     }
52     return double(count) / runs;
53 }
54
55 double average_len_after_n(int n, int runs) {
56     long long count = 0;

```

```

57     for (int i = 0; i < runs; i++) {
58         count += annihilate(gen_tube(n, 0.5)).length();
59     }
60     return count / double(runs);
61 }
62
63 double threaded_avg_len_after_n(int n, int runs) {
64     double avg = 0;
65     vector<future<double>> threads;
66     for (int i = 0; i < num_threads; i++) {
67         threads.push_back(
68             async(launch::async, average_len_after_n, n, runs / num_threads));
69     }
70     for (auto &t : threads) {
71         avg += t.get();
72     }
73     return avg / num_threads;
74 }
75
76 double proportion_x(int runs, double p) {
77     string tube = "";
78     vector<future<string>> threads;
79     for (int i = 0; i < num_threads; i++) {
80         string small_tube = gen_tube(runs / num_threads, p);
81         threads.push_back(async(launch::async, annihilate, small_tube));
82     }
83     for (auto &t : threads) {
84         tube += t.get();
85     }
86     int len = tube.length();
87     return (len - (1 - p) * runs) / len;
88 }
89
90 int main() {
91
92     long long runs = 1000000;
93
94     cout << "q1\n";
95     for (int i = 1; i <= 10; i++) {
96         cout << "n = " << i << " prob = " << prob_no_two_consec_after_n(i, runs)
97             << '\n';
98     }
99     cout << "q2\n";
100    for (int i = 1; i <= 10; i++) {
101        cout << "n = " << i << " avg length = " << threaded_avg_len_after_n(i, runs)
102            << '\n';
103    }
104    cout << "q3\n";
105    for (double p = 1; p < 10; p++) {
106        cout << "p = " << p / 10 << " proportion = " << proportion_x(runs, p / 10)
107            << '\n';
108    }
109
110    return 0;
111 }

```

When compiled and executed, this code outputs the following. Note that the code uses a random number generator, meaning it will not output exact values.

```

q1
n = 1 prob = 1
n = 2 prob = 0.750638

```

```

n = 3 prob = 0.624614
n = 4 prob = 0.500102
n = 5 prob = 0.406429
n = 6 prob = 0.32788
n = 7 prob = 0.265315
n = 8 prob = 0.215035
n = 9 prob = 0.17349
n = 10 prob = 0.140863
q2
n = 1 avg length = 1
n = 2 avg length = 1.74977
n = 3 avg length = 2.49675
n = 4 avg length = 3.24339
n = 5 avg length = 3.99241
n = 6 avg length = 4.73593
n = 7 avg length = 5.48245
n = 8 avg length = 6.23626
n = 9 avg length = 6.9817
n = 10 avg length = 7.73298
q3
p = 0.1 proportion = 0.0909844
p = 0.2 proportion = 0.166457
p = 0.3 proportion = 0.231049
p = 0.4 proportion = 0.285553
p = 0.5 proportion = 0.333787
p = 0.6 proportion = 0.374153
p = 0.7 proportion = 0.411355
p = 0.8 proportion = 0.446417
p = 0.9 proportion = 0.471816

```

This output matches with the theoretical values using formulas obtained earlier.

CODE IMPLEMENTATION FOR GENERALISATION

```

1  #include <future>
2  #include <iostream>
3  #include <random>
4  #include <string>
5  #include <thread>
6  #include <vector>
7
8  using namespace std;
9  using namespace std::chrono;
10
11 random_device rd;
12
13 const char particles[] = "XY";
14 unsigned int num_threads = thread::hardware_concurrency();
15
16 using namespace std;
17
18 string gen_tube(int length, double p) {
19     mt19937 rng(rd());
20     discrete_distribution<int> pick{p, 1 - p};
21     string tube;
22     for (int i = 0; i < length; i++) {
23         tube += particles[pick(rng)];
24     }
25     return tube;
26 }

```



```

27
28 string binary_to_tube(string s) {
29     string out;
30     for (char c : s)
31         (c == '1') ? (out += 'X') : (out += 'Y');
32     return out;
33 }
34
35 string annihilate(string A, int m, int n) {
36     long long cnt = 0;
37     string B;
38     for (char x : A) {
39         long long k = cnt;
40         while (k >= m)
41             k = (k / m) * n + (k % m);
42         x == 'X' ? (cnt++) : (B += string(k, 'X') + "Y", cnt = 0);
43     }
44     while (cnt >= m)
45         cnt = (cnt / m) * n + (cnt % m);
46     B += string(cnt, 'X');
47     return B;
48 }
49
50 long double average_len_after_n(int k, int runs, int m, int n, double p) {
51     unsigned long long count = 0;
52     for (int i = 0; i < runs; i++) {
53         count += annihilate(gen_tube(k, p), m, n).length();
54     }
55     return count / double(runs);
56 }
57
58 long double brute_force_avg(int k, int m, int n) {
59     long long cnt = 0;
60     for (int i = 0; i < (1 << k); i++) {
61         string tube;
62         for (int j = k - 1; j >= 0; j--) {
63             tube += to_string((i >> j) & 1);
64         }
65         cnt += annihilate(binary_to_tube(tube), m, n).length();
66     }
67     return (long double)(cnt) / (1 << k);
68 }
69
70 long double threaded_avg_len_after_n(int k, int runs, int m, int n, double p) {
71     long double avg = 0;
72     vector<future<long double>> threads;
73     for (int i = 0; i < num_threads; i++) {
74         threads.push_back(async(launch::async, average_len_after_n, k,
75                                 runs / num_threads, m, n, p));
76     }
77     for (auto &t : threads) {
78         avg += t.get();
79     }
80     return avg / num_threads;
81 }
82
83 int main() {
84     long long runs = 1000000;
85
86     int m = 6;
87     int n = 5;
88

```

```

89
90     for (int i = 1; i <= 15; i++) {
91         cout << "n = " << i << " m = " << m << " n = " << n
92             << " p = 0.5 avg = " << brute_force_avg(i, m, n) << '\n';
93     }
94
95     m = 4;
96     n = 1;
97
98     double p = 0.42;
99     for (int i = 1; i <= 15; i++) {
100         cout << "n = " << i << " m = " << m << " n = " << n << " p = " << p
101             << " avg = " << threaded_avg_len_after_n(i, runs, m, n, p) << '\n';
102     }
103
104     return 0;
105 }

```

When the $p = \frac{1}{2}$, the program will output exact values (rounded by C++). Otherwise, the program uses a random number generator.

```

n = 1 m = 6 n = 5 p = 0.5 avg = 1
n = 2 m = 6 n = 5 p = 0.5 avg = 2
n = 3 m = 6 n = 5 p = 0.5 avg = 3
n = 4 m = 6 n = 5 p = 0.5 avg = 4
n = 5 m = 6 n = 5 p = 0.5 avg = 5
n = 6 m = 6 n = 5 p = 0.5 avg = 5.98438
n = 7 m = 6 n = 5 p = 0.5 avg = 6.96875
n = 8 m = 6 n = 5 p = 0.5 avg = 7.95312
n = 9 m = 6 n = 5 p = 0.5 avg = 8.9375
n = 10 m = 6 n = 5 p = 0.5 avg = 9.92188
n = 11 m = 6 n = 5 p = 0.5 avg = 10.9062
n = 12 m = 6 n = 5 p = 0.5 avg = 11.8906
n = 13 m = 6 n = 5 p = 0.5 avg = 12.875
n = 14 m = 6 n = 5 p = 0.5 avg = 13.8594
n = 15 m = 6 n = 5 p = 0.5 avg = 14.8438
n = 1 m = 4 n = 1 p = 0.42 avg = 1
n = 2 m = 4 n = 1 p = 0.42 avg = 2
n = 3 m = 4 n = 1 p = 0.42 avg = 3
n = 4 m = 4 n = 1 p = 0.42 avg = 3.90659
n = 5 m = 4 n = 1 p = 0.42 avg = 4.85173
n = 6 m = 4 n = 1 p = 0.42 avg = 5.79867
n = 7 m = 4 n = 1 p = 0.42 avg = 6.73798
n = 8 m = 4 n = 1 p = 0.42 avg = 7.67977
n = 9 m = 4 n = 1 p = 0.42 avg = 8.6216
n = 10 m = 4 n = 1 p = 0.42 avg = 9.56162
n = 11 m = 4 n = 1 p = 0.42 avg = 10.5018
n = 12 m = 4 n = 1 p = 0.42 avg = 11.4467
n = 13 m = 4 n = 1 p = 0.42 avg = 12.3881
n = 14 m = 4 n = 1 p = 0.42 avg = 13.3287
n = 15 m = 4 n = 1 p = 0.42 avg = 14.2699

```

REFERENCES

- [1] Arthur T. Benjamin and Jennifer J. Quinn. *Proofs that Really Count: The Art of Combinatorial Proof*. The Dolciani Mathematical Expositions. The Mathematical Association of America, 2003.