

# Putnam 2024 Problems

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December 2024

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## 1 Problem A5

Consider a circle  $\Omega$  with radius 9 and center at the origin  $(0, 0)$ , and a disk  $\Delta$  with radius 1 and center at  $(r, 0)$ , where  $0 \leq r \leq 8$ . Two points  $P$  and  $Q$  are chosen independently and uniformly at random on  $\Omega$ . Which value(s) of  $r$  minimize the probability that the chord  $\overline{PQ}$  intersects  $\Delta$ ?

### 1.1 Solution

Let the two random points on  $\Omega$  be  $P(9 \cos \theta_1, 9 \sin \theta_1)$  and  $Q(9 \cos \theta_2, 9 \sin \theta_2)$  where  $0 \leq \theta_1, \theta_2 \leq 2\pi$ . The slope of the chord  $\overline{PQ}$  is

$$\frac{9 \sin \theta_2 - 9 \sin \theta_1}{9 \cos \theta_2 - 9 \cos \theta_1} = -\cot\left(\frac{\theta_1 + \theta_2}{2}\right). \quad (1.1)$$

The equation of the line passing through  $P$  and  $Q$  is given by

$$(y - 9 \sin \theta_1) = -\cot\left(\frac{\theta_1 + \theta_2}{2}\right)(x - 9 \cos \theta_1). \quad (1.2)$$

The perpendicular distance  $d$  of the point from the center of disk  $\Delta$  at  $(r, 0)$  to the above line is given by:

$$\begin{aligned} d &= \frac{\left| (r - 9 \cos \theta_1) \cot\left(\frac{\theta_1 + \theta_2}{2}\right) - 9 \sin \theta_1 \right|}{\sqrt{1 + \cot^2\left(\frac{\theta_1 + \theta_2}{2}\right)}} \\ &= \left| \sin\left(\frac{\theta_1 + \theta_2}{2}\right) \left| r \cot\left(\frac{\theta_1 + \theta_2}{2}\right) - 9 \csc\left(\frac{\theta_1 + \theta_2}{2}\right) \cos\left(\frac{\theta_1 - \theta_2}{2}\right) \right| \right| \\ &= \left| r \cos\left(\frac{\theta_1 + \theta_2}{2}\right) - 9 \cos\left(\frac{\theta_1 - \theta_2}{2}\right) \right|. \end{aligned} \quad (1.3)$$

For the chord  $\overline{PQ}$  to intersect  $\Delta$ ,  $d \leq 1$ , i.e.

$$-1 \leq r \cos\left(\frac{\theta_1 + \theta_2}{2}\right) - 9 \cos\left(\frac{\theta_1 - \theta_2}{2}\right) \leq 1. \quad (1.4)$$

Let  $A(r, \theta_1, \theta_2)$  denote the area of the region given by Equation (1.4), the probability of the chord  $\overline{PQ}$  intersecting  $\Delta$  as a function of  $r$  is given by

$$P(r) = \frac{A(r, \theta_1, \theta_2)}{2\pi \cdot 2\pi} = \frac{A(r, \theta_1, \theta_2)}{4\pi^2}. \quad (1.5)$$

To be able to recast  $A(r, \theta_1, \theta_2)$  as an integral, we use the substitution  $\frac{\theta_1 + \theta_2}{2} = u$  and  $\frac{\theta_1 - \theta_2}{2} = v$ . We now have  $0 \leq u \leq 2\pi$ ,  $-\pi \leq v \leq \pi$  and the inequality Equation (1.4) is transformed into

$$-1 \leq r \cos u - 9 \cos v \leq 1. \quad (1.6)$$

As  $\theta_1 = u + v$ ,  $\theta_2 = u - v$ , the Jacobian matrix is

$$J = \begin{pmatrix} \frac{\partial \theta_1}{\partial u} & \frac{\partial \theta_1}{\partial v} \\ \frac{\partial \theta_2}{\partial u} & \frac{\partial \theta_2}{\partial v} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}. \quad (1.7)$$

Therefore,  $|\det(J)|$  for the transformation is 2. If we denote the region given by Equation (1.6) as  $S(r, u, v)$ , the probability of the chord  $\overline{PQ}$  intersecting  $\Delta$  as a function of  $r$  is given by

$$P(r) = \frac{S(r, u, v)}{2 \cdot 2\pi \cdot 2\pi} = \frac{S(r, u, v)}{8\pi^2}. \quad (1.8)$$

From Equation (1.6), we get

$$\begin{aligned} \frac{r \cos u + 1}{9} \geq \cos v \geq \frac{r \cos u - 1}{9} \\ \Rightarrow \arccos\left(\frac{r \cos u - 1}{9}\right) \geq v \geq \arccos\left(\frac{r \cos u + 1}{9}\right), 0 \leq v \leq \pi \end{aligned} \quad (1.9)$$

as the cosine function is one-to-one and decreasing in  $[0, \pi]$ . We can now write  $S(r, u, v)$  as an integral,

$$S(r, u, v) = 2 \int_0^{2\pi} \int_{\arccos\left(\frac{r \cos u + 1}{9}\right)}^{\arccos\left(\frac{r \cos u - 1}{9}\right)} 2dvdu = 16 \int_0^{\frac{\pi}{2}} g(r, u)du \quad (1.10)$$

where

$$g(r, u) = \arccos\left(\frac{r \cos u - 1}{9}\right) - \arccos\left(\frac{r \cos u + 1}{9}\right). \quad (1.11)$$

We have

$$P(r) = \frac{2}{\pi^2} \int_0^{\frac{\pi}{2}} g(r, u)du \Rightarrow P'(r) = \frac{2}{\pi^2} \int_0^{\frac{\pi}{2}} \frac{\partial g(r, u)}{\partial r} du. \quad (1.12)$$

Now

$$\frac{\partial g(r, u)}{\partial r} = (\cos u) \left( (80 - 2r \cos u - r^2 \cos^2 u)^{-\frac{1}{2}} - (80 + 2r \cos u - r^2 \cos^2 u)^{-\frac{1}{2}} \right) \quad (1.13)$$

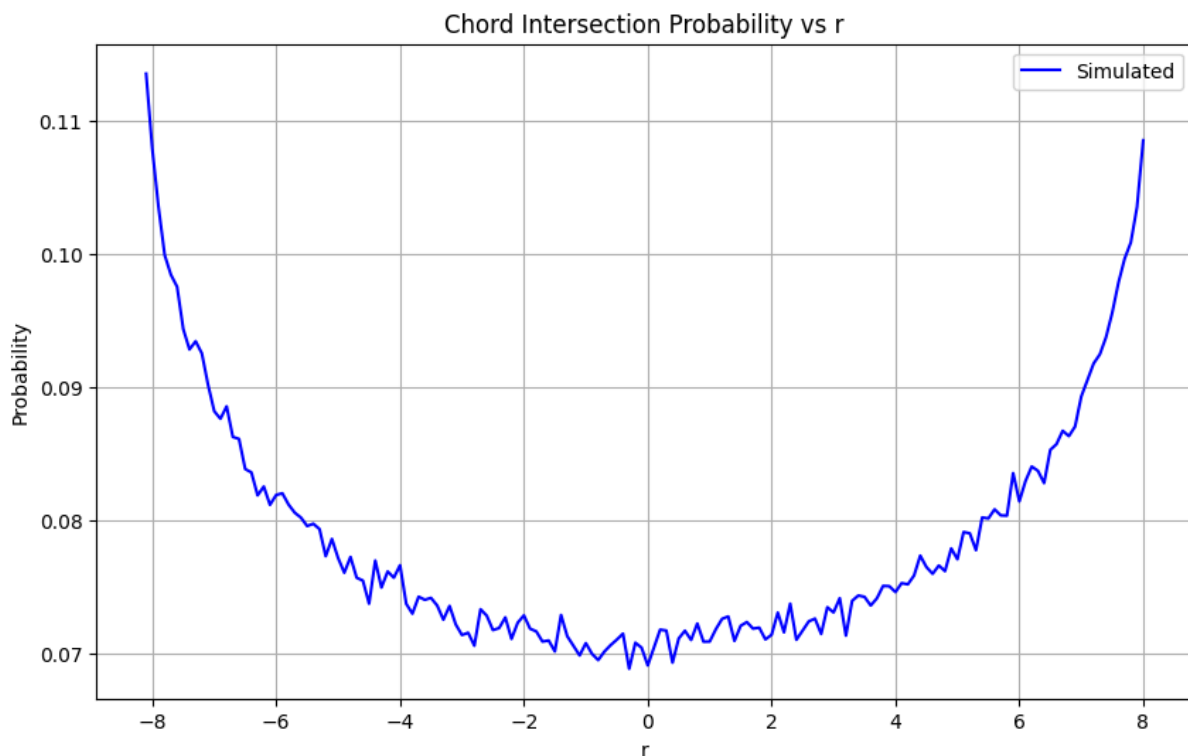
which, for  $u \in \left(0, \frac{\pi}{2}\right)$ , is zero for  $r = 0$  and strictly positive for  $r > 0$ . It follows that  $P'(0) = 0$  and  $P'(r) > 0$  for  $r \in (0, 8]$ . Therefore,  $P(r)$  is minimized for  $\mathbf{r} = \mathbf{0}$ .

## 1.2 Estimation of probability using Monte-Carlo

In the second approach, we follow the procedure below to estimate the probability of intersection:

- (i) Choose two random points on the circumference of  $\Omega$ .
- (ii) Derive the equation of the line passing through the two points chosen above.
- (iii) Calculate the perpendicular distance from the center  $(r, 0)$  of  $\Delta$  to the line using the formula  $\left| \frac{ar+c}{\sqrt{a^2+b^2}} \right|$  where  $ax + by + c = 0$  is the equation of the line.
- (iv) Repeat the above steps multiple times and calculate the fraction of trials where the perpendicular distance is less than 1, the radius of  $\Delta$ .

The above procedure gives us the following plot of the estimate of the probability as a function of  $r$ :



From the symmetry of the plot we can see that the probability is minimized when  $r = 0$ .

### 1.2.1 Python code

```
import numpy as np
import matplotlib.pyplot as plt
from tqdm import tqdm

def get_line_equation(p1, p2):
    """Returns a, b, c for line ax + by + c = 0"""
    a = p2[1] - p1[1]
    b = p1[0] - p2[0]
    c = p2[0]*p1[1] - p1[0]*p2[1]
    return a, b, c

def perpendicular_distance(line_coeffs, point):
    """Calculate perpendicular distance from point to line ax + by + c = 0"""
    a, b, c = line_coeffs
```

```

x0, y0 = point
return abs(a*x0 + b*y0 + c) / np.sqrt(a*a + b*b)

def simulate_probability(r, n_trials=100000):
    """Simulate using perpendicular distance method"""
    R = 9 # radius of large circle
    intersections = 0

    for _ in range(n_trials):
        # Generate random points on circle
        theta1, theta2 = np.random.uniform(0, 2*np.pi, 2)
        p1 = R * np.array([np.cos(theta1), np.sin(theta1)])
        p2 = R * np.array([np.cos(theta2), np.sin(theta2)])

        # Get line equation and calculate distance
        line_coeffs = get_line_equation(p1, p2)
        dist = perpendicular_distance(line_coeffs, [r, 0])

        if dist <= 1: # radius of small circle is 1
            intersections += 1

    return intersections / n_trials

# Run simulation
r_values = np.arange(-8.1, 8.1, 0.1)
probabilities = []

for r in tqdm(r_values):
    prob = simulate_probability(r)
    probabilities.append(prob)

# Plot results
plt.figure(figsize=(10, 6))
plt.plot(r_values, probabilities, 'b-', label='Simulated')
plt.grid(True)
plt.xlabel('r')
plt.ylabel('Probability')
plt.title('Chord Intersection Probability vs r')
plt.legend()
plt.show()

```

## 2 Problem B4

Let  $n$  be a positive integer. Set  $a_{n,0} = 1$ . For  $k \geq 0$ , choose an integer  $m_{n,k}$  uniformly at random from the set  $\{1, 2, \dots, n\}$ , and let

$$a_{n,k+1} = \begin{cases} a_{n,k} + 1, & \text{if } m_{n,k} > a_{n,k}; \\ a_{n,k}, & \text{if } m_{n,k} = a_{n,k}; \\ a_{n,k} - 1, & \text{if } m_{n,k} < a_{n,k}. \end{cases} \quad (2.14)$$

Let  $E(n)$  be the expected value of  $a_{n,n}$ . Determine  $\lim_{n \rightarrow \infty} \frac{E(n)}{n}$ .

### 2.1 Solution

Firstly, we notice that  $1 \leq a_{n,n} \leq n$ . Dropping the subscript  $n$  we can write the equation Equation (2.14) as

$$a_{k+1} = a_k + X \quad (2.15)$$

where  $X$  is a random variable that takes the values 1, 0 and  $-1$  when  $m_k > a_k$ ,  $m_k = a_k$  and  $m_k < a_k$  respectively. Taking expectation on both sides of Equation (2.15) we get

$$\mathbb{E}[a_{k+1}] = \mathbb{E}[a_k] + \mathbb{E}[X]. \quad (2.16)$$

To calculate  $\mathbb{E}[X]$ , we make use of the **Law of Iterated Expectations**,

$$\mathbb{E}[X] = \mathbb{E}[\mathbb{E}[X|a_k]] = \mathbb{E}\left[1 \cdot \frac{n - a_k}{n} + (-1) \cdot \frac{a_k - 1}{n}\right] = \left(-\frac{2}{n}\right)\mathbb{E}[a_k] + \frac{n+1}{n}. \quad (2.17)$$

Substituting the value of  $\mathbb{E}[X]$  from Equation (2.17) in Equation (2.16), we get

$$\mathbb{E}[a_{k+1}] = \frac{n-2}{n}\mathbb{E}[a_k] + \frac{n+1}{n}. \quad (2.18)$$

Setting  $b(k) = \mathbb{E}[a_k]$  in Equation (2.18), we get the following recurrence relation

$$b(k+1) = \frac{n-2}{n}b(k) + \frac{n+1}{n} \quad (2.19)$$

with  $b(0) = 1$ . We use generating functions to solve the above linear recurrence relation. Let

$$G(z) = \sum_{k=0}^{\infty} b(k)z^k \quad (2.20)$$

Multiplying both sides of the recurrence relation by  $z^k$  and summing up both sides of the equation term by term over the nonnegative integers, we get

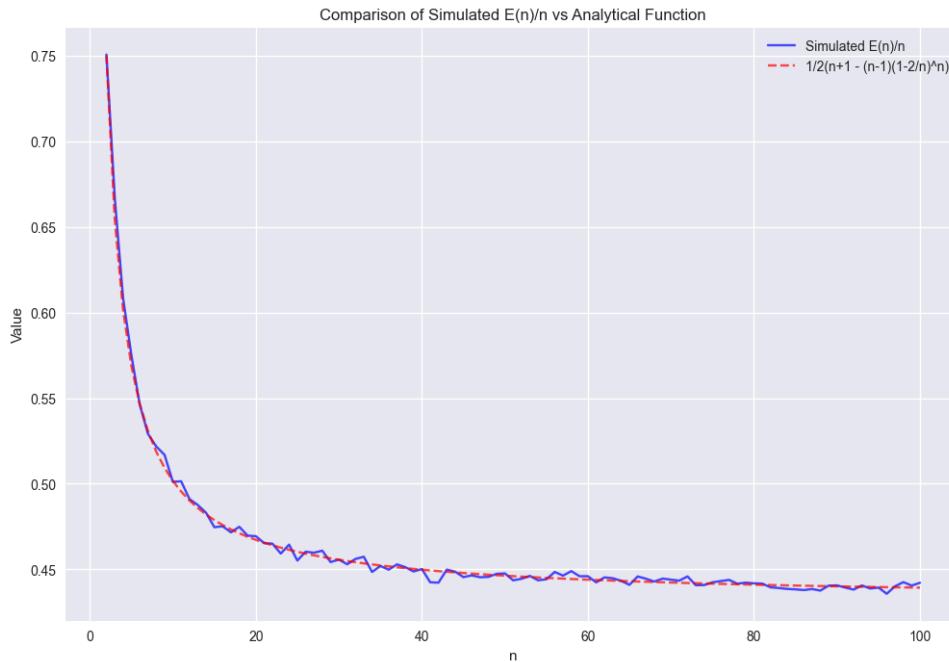
$$\begin{aligned}
& \sum_{k=0}^{\infty} b(k+1)z^k + \frac{2-n}{n} \sum_{k=0}^{\infty} b(k)z^k = \frac{n+1}{n} \sum_{k=0}^{\infty} z^k \\
\Rightarrow & \sum_{k=0}^{\infty} b(k+1)z^{k+1}z^{-1} + \frac{2-n}{n}G(z) = \frac{n+1}{n(1-z)} \\
\Rightarrow & (G(z) - b(0))z^{-1} + \frac{2-n}{n}G(z) = \frac{n+1}{n(1-z)} \\
\Rightarrow & G(z) = \frac{n+z}{(n-(n-2)z)(1-z)} \\
\Rightarrow & G(z) = \frac{1}{2} \sum_{k=0}^{\infty} \left( n+1 - (n-1) \left( \frac{n-2}{n} \right)^k \right) z^k.
\end{aligned} \tag{2.21}$$

Therefore,

$$\lim_{n \rightarrow \infty} \frac{E(n)}{n} = \lim_{n \rightarrow \infty} \frac{b(n)}{n} = \lim_{n \rightarrow \infty} \frac{1}{2} \frac{n+1 - (n-1) \left( \frac{n-2}{n} \right)^n}{n} = \frac{1}{2} (1 - e^{-2}). \tag{2.22}$$

## 2.2 Computational verification

From the code below we get the following plot for the simulated value of  $\frac{E(n)}{n}$  against  $\frac{b(n)}{n}$  calculated analytically:



## 2.2.1 Python Code

```
import numpy as np
import matplotlib.pyplot as plt

def simulate_process(n, num_trials=1000):
    final_values = []
    for _ in range(num_trials):
        a = 1
        for _ in range(n):
            m = np.random.randint(1, n+1)
            if m > a:
                a += 1
            elif m < a:
                a -= 1
        final_values.append(a)
    return np.mean(final_values)

def analytical_function(n):
    return 0.5 * ((n+1) - (n-1)*((n-2)/n)**n)/n

# Generate data points
n_values = np.arange(2, 101) # n from 1 to 100
simulated_values = []
analytical_values = []

# Calculate both simulated and analytical values
print("Running simulation...")
for n in n_values:
    if n == 1: # Special case for n=1
        simulated_values.append(1)
    else:
        en = simulate_process(n)
        simulated_values.append(en/n)
        analytical_values.append(analytical_function(n))

# Create the plot
plt.figure(figsize=(12, 8))
plt.plot(n_values, simulated_values, 'b-', label='Simulated E(n)/n', alpha=0.7)
plt.plot(n_values, analytical_values, 'r--', label='1/2(n+1 - (n-1)(1-2/n)^n)',
alpha=0.7)
plt.xlabel('n')
plt.ylabel('Value')
plt.title('Comparison of Simulated E(n)/n vs Analytical Function')
plt.grid(True)
plt.legend()
plt.show()
```