

①

Favorable events

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Case#	C ₁	C ₂
1	H T	T T
2	T H	T T
3	H H	T T
4	H H	H T
5	H H	T H

$$P(\text{case \#1}) = P(\text{case \#2}) = \frac{1}{2} \times \frac{1}{2} \times \frac{1}{3} \times \frac{1}{3} \\ = \frac{1}{36}$$

$$P(\text{case \#3}) = \frac{1}{2} \times \frac{1}{2} \times \frac{1}{3} \times \frac{1}{3} = \frac{1}{36}$$

$$P(\text{case \#4}) = P(\text{case \#5}) = \frac{1}{2} \times \frac{1}{2} \times \frac{2}{3} \times \frac{1}{3} = \frac{2}{36}$$

$$P(X > Y) = \sum_{i=1}^5 P(\text{case \# } i) = \frac{7}{36}$$

② $P(A) = 0.7$ $P(B) = 0.9$

$$P(A \cup B) - P(A \cap B) = P(A) + P(B) - 2 P(A \cap B)$$

$$\text{Max } (P(A \cup B) - P(A \cap B)) = P(A) + P(B) - 2 \min(P(A \cap B))$$

Minimum occurs for independent events

$$= P(A) + P(B) - 2 P(A) \times P(B)$$

$$= 0.7 + 0.9 - 2 \times 0.7 \times 0.9$$

$$= \boxed{0.34}$$

$$\text{Min } (P(A \cup B) - P(A \cap B)) = P(A) + P(B) - 2 \max(P(A \cap B))$$

Maximum occurs when least probable event is inside most probable event

$$= P(A) + P(B) - 2 P(A)$$

$$= P(B) - P(A)$$

$$= 0.9 - 0.7 = \boxed{0.2}$$

③

H_1

H_2

H_3

H_4

H_5



P_1



P_2



P_3

$$P(\text{exactly two in same Hotel}) = P(P_1, P_2 \text{ in same hotel and } P_3 \text{ in different})$$

$$+ P(P_2, P_3 \text{ in same hotel and } P_1 \text{ in different})$$

$$+ P(P_3, P_1 \text{ in same hotel and } P_2 \text{ in different})$$

$$= 3 \times \left[\frac{5 \times 4}{5 \times 5 \times 5} \right] \text{ Total number of possible ways}$$

$$= \frac{12}{25}$$

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Geometric Distribution

$$P(X=n) = (1-p)^{n-1} p, n=1, 2, \dots$$

$$E(X) = \sum_{i=1}^{\infty} i \cdot P(X=i) = \sum_{i=1}^{\infty} i (1-p)^{i-1} p$$

$$= p \times \sum_{i=1}^{\infty} -\frac{d}{dp} [(1-p)^i]$$

$$= -p \times \frac{d}{dp} \sum_{i=1}^{\infty} (1-p)^i$$

$$= -p \times \frac{d}{dp} [(1-p) + (1-p)^2 + \dots \infty]$$

$$= -p \times \frac{d}{dp} \left(\frac{(1-p)}{1-(1-p)} \right)$$

$$= -p \times \frac{d}{dp} \left(\frac{(1-p)}{p} \right)$$

$$= -p \times \frac{[p(-1) - (1-p)]}{p^2}$$

$$= -\left[\frac{-p-1+p}{p} \right] = \boxed{+\frac{1}{p}}$$

$$\text{Var}[x] = E[x^2] - (E[x])^2$$

$$E[x^2] = \sum_{i=1}^{\infty} i^2 (1-p)^{i-1} p = p \sum_{i=1}^{\infty} i^2 (1-p)^{i-1}$$

$$= p \cdot \frac{1 + (1-p)}{[1 - (1-p)]^2} = \frac{(2-p)}{p^2}$$

$$\text{Var}[x] = \frac{2-p}{p^2} - \frac{1}{p^2}$$

$$= \boxed{\frac{1-p}{p^2}}$$

⑤

$$\text{R.H.S.} = \sum_{n=1}^{\infty} P(N \geq n)$$

$$= P(N \geq 1) + P(N \geq 2) + P(N \geq 3) + \dots \infty$$

$$= \left[\underline{P(N=1)} + \underline{P(N=2)} + \underline{P(N=3)} + \dots \infty \right] +$$
$$\left[\underline{P(N=2)} + \underline{P(N=3)} + P(N=4) + \dots \infty \right] +$$
$$\left[\underline{P(N=3)} + P(N=4) + P(N=5) + \dots \infty \right]$$

$$= 1 \times P(N=1) + 2 \times P(N=2) + 3 \times P(N=3) + \dots \infty$$

$$= \sum_{i=1}^{\infty} i P(N=i) = E(N) = \text{L.H.S.}$$

Hence, proved.

⑥

$$x_1 + x_2 + x_3 + \dots + x_n = r$$

$$\left. \begin{array}{l} \text{Total number of ways} \\ \text{to distribute } r \text{ balls in} \\ n \text{ boxes} \end{array} \right\} = {}^{n+r-1}C_{n-1}$$

Case #1: First box is empty

$$\left. \begin{array}{l} \text{Total number of ways to} \\ \text{distribute } r \text{ balls in} \\ \text{boxes labeled 2 to } n \end{array} \right\} = {}^{n+r-2}C_{n-2}$$

$$\text{Probability} = \frac{{}^{n+r-2}C_{n-2}}{{}^{n+r-1}C_{n-1}} = \frac{(n+r-2)! \times r! \times (n-1)!}{r! \times (n-2)! \times (n+r-1)!}$$

$$= \frac{(n-1)}{(n+r-1)}$$

Case #2: First two boxes are empty

$$\left. \begin{array}{l} \text{Total number of ways to} \\ \text{distribute } r \text{ balls in boxes} \\ \text{labelled 3 to } n \end{array} \right\} = {}^{n+r-3}C_{n-3}$$

$$\text{Probability} = \frac{{}^{n+r-3}C_{n-3}}{{}^{n+r-1}C_{n-1}} = \frac{(n+r-3)! \times r! \times (n-1)!}{r! \times (n-3)! \times (n+r-1)!}$$

$$= \frac{(n-1)(n-2)}{(n+r-1)(n+r-2)}$$

⑦

$$f(x) = \begin{cases} 0 & , x < 1 \\ e^{-(x-1)} & , x \geq 1 \end{cases}$$

$$Y = x^2 \Rightarrow Y \in [1, \infty)$$

C.D.F. $F_Y(y)$ for $y \in [1, \infty)$

$$F_Y(y) = P(Y \leq y)$$

$$= P(x^2 \leq y)$$

$$= P(-\sqrt{y} \leq x \leq \sqrt{y})$$

$$= \underbrace{P(-\sqrt{y} \leq x \leq 1)}_0 + P(1 \leq x \leq \sqrt{y})$$

$$= \int_1^{\sqrt{y}} e^{-(x-1)} dx = 1 - e^{(1-\sqrt{y})}$$

$$\therefore \text{C.D.F. of } Y, F_Y(y) = \begin{cases} 0, & y < 1 \\ 1 - e^{(1-\sqrt{y})}, & y \geq 1 \end{cases}$$

$$\text{P.D.F. of } Y = f_Y(y) = F'_Y(y) = \begin{cases} 0, & y < 1 \\ \frac{e^{(1-\sqrt{y})}}{2\sqrt{y}}, & y \geq 1 \end{cases}$$

$$\begin{aligned}
 \textcircled{8} \quad \text{Var}[X_1 + X_2 + \dots + X_n] &= E \left[(X_1 + X_2 + \dots + X_n) - E(X_1 + X_2 + \dots + X_n) \right]^2 \\
 &= E \left[(X_1 + X_2 + \dots + X_n) - (E(X_1) + E(X_2) + \dots + E(X_n)) \right]^2 \\
 &= E \left[\underbrace{(X_1 - E(X_1)) + (X_2 - E(X_2)) + \dots + (X_n - E(X_n))}_{(1)} \right]^2
 \end{aligned}$$

From Algebra, we know that,

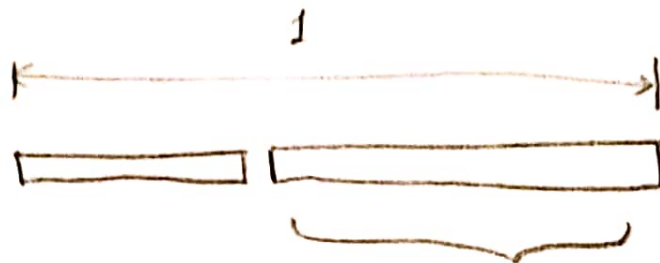
$$\boxed{\sum_{i=1}^N a_i^2 + 2 \sum_{i=1}^N \sum_{j=1}^{i-1} a_i a_j = \left(\sum_{i=1}^N a_i \right)^2}$$

Therefore, (1) becomes,

$$\begin{aligned}
 &= E \left[\sum_{i=1}^n (X_i - E(X_i))^2 + 2 \sum_{j < i} (X_i - E(X_i))(X_j - E(X_j)) \right] \\
 &= \sum_{i=1}^n E(X_i - E(X_i))^2 + 2 \sum_{j < i} E(X_i - E(X_i))(X_j - E(X_j)) \\
 &= \sum_{i=1}^n \text{Var}[X_i] + 2 \sum_{j < i} \text{COV}(X_i, X_j)
 \end{aligned}$$

Hence, Proved

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$$Y = \begin{cases} 1-x & \text{if } x < \frac{1}{2} \\ x & \text{if } x \geq \frac{1}{2} \end{cases}$$

$$E(Y) = \int_0^{0.5} (1-x) dx + \int_{0.5}^1 x dx$$

$$= \left[x - \frac{x^2}{2} \right]_0^{0.5} + \left[\frac{x^2}{2} \right]_{0.5}^1 = \boxed{\frac{3}{4}}$$

$$\text{Var}(Y) = \int_0^{0.5} (1-x)^2 dx + \int_{0.5}^1 x^2 dx$$
$$= \left[-\frac{(1-x)^3}{3} \right]_0^{0.5} + \left[\frac{x^3}{3} \right]_{0.5}^1$$

$$= \frac{(1 - 0.125) + (1 - 0.125)}{3}$$

$$= \boxed{\frac{7}{12}}$$

10)

Number of runs = 10^6

Language : Python

Method : Monte Carlo Simulation

Volume under
curve $z = \frac{1}{3} \cosh \sqrt{x^2 + y^2}$ } = $P(\text{Success}) \times \text{Volume of Cube}$

$$= \frac{\text{inside_points}}{\text{total_points}} \times 1$$

$$\approx \boxed{0.45}$$

(11)

"n" Red ; "m" Black

(i) replacement

$$X = R_1 + R_2 + \dots + R_k$$

$$E[X] = E[R_1 + R_2 + \dots + R_k]$$

$$i = 1, \dots, k$$

$$R_i = \begin{cases} 1 & \text{if the } i\text{th ball is Red} \\ 0 & \text{if the } i\text{th ball is Black} \end{cases}$$

$$P(R_i) = \frac{n}{m+n}$$

$$E[X] = \frac{kn}{m+n}$$

$$\begin{aligned} \text{Var}[X] &= \text{Var}[R_1 + R_2 + \dots + R_k] = \sum_{i=1}^k \text{Var}(R_i) + 2 \sum_{p < q} \underbrace{\text{Cov}(R_p, R_q)}_{\substack{0 \\ \text{independent events}}} \\ &= \left[\text{Var}(R_1) + \text{Var}(R_2) + \dots + \text{Var}(R_k) \right] \end{aligned}$$

$$= k \times \text{Var}(R_1)$$

$$= k \times \left(E(R_1^2) - [E(R_1)]^2 \right)$$

$$= k \times \left(1^2 \times \frac{n}{m+n} - \frac{n^2}{(m+n)^2} \right)$$

$$= k \left(\frac{nm + n^2 - n^2}{(m+n)^2} \right) = \boxed{\frac{kmn}{(m+n)^2}}$$

(ii) without replacement

$$E[X] = \sum_{i=1}^k E[R_i]$$

$$E[R_i] = \frac{n}{m+n}$$

$$\therefore E[X] = \frac{kn}{m+n}$$

$$\text{Var}[X] = \text{Var}[R_1 + R_2 + \dots + R_k]$$

$$= \sum_{i=1}^k \text{Var}[R_i] + 2 \sum_{1 \leq i < j \leq k} \text{Cov}(R_i, R_j)$$

↑
all same
for all i

$$\text{Var}[R_i] = E[R_i^2] - (E[R_i])^2$$

$$= 1^2 \times \frac{n}{m+n} - \left(\frac{n}{m+n}\right)^2 = \frac{mn}{(m+n)^2}$$

$$\text{Cov}(R_i, R_j) = P(R_i=1, R_j=1) - P(R_i=1)P(R_j=1)$$

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