

$$1. \quad P(\text{at least 1 correct}) = 1 - \frac{D(N)}{N!}$$

Derangement's formula,

$$D(N) = N! \left(1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \dots + \frac{(-1)^N}{N!} \right)$$

probability of all letters misplaced,

$$\frac{D(N)}{N!} = \sum_{k=0}^N \frac{(-1)^k}{k!}$$

$$\text{as } N \rightarrow \infty, \quad \sum_{k=0}^{\infty} \frac{(-1)^k}{k!} = e^{-1}$$

for $N = 50$, error term $\frac{(-1)^{51}}{51!}$ is negligible,

$$\frac{D(50)}{50!} \approx \frac{1}{e}$$

$$P(\text{at least one correct}) \approx 1 - \frac{1}{e}$$

$$P \approx 1 - \frac{1}{e} \approx 0.6321$$

2. Initially, it is equally likely that \$1000 present in any present.

$P_1 : \frac{1}{3}$
 $P_2 : \frac{1}{3}$
 $P_3 : \frac{1}{3}$

} Probability of good gift

Case 1: \$1000 in P_1 (you picked it)

— Probability = $\frac{1}{3}$.

host opens 2 or 3 with equal probability $\Rightarrow \frac{1}{2}$.

Given, host opened P_2 with $P = \frac{1}{2}$.

Case 2: \$1000 in $P_2 \Rightarrow P = \frac{1}{3}$

• Host cannot open P_2 .

must open P_3

since, you picked P_1 , host must open P_3 .

This case is ruled out (since 2 is opened).

Case 3: if \$ 1000 in P_3 .

host must open P_2 .

$$\underline{\underline{P = 1}},$$

• Conditional prob. (Picked 1 & opened 2)

$$P(G_1 \cap P_2) = \frac{1}{3} \times \frac{1}{2} = \frac{1}{6}.$$

\downarrow \downarrow
money in opened P_2
 P_1

$$P(G_3 \cap P_2) = \frac{1}{3} \times 1 = \frac{1}{3}.$$

$$P(G_2 \cap P_2) = 0$$

$$P(\text{host opens } P_2) = \frac{1}{2}.$$

$$\text{So, } P(G_1 \mid \text{host opens } 2) = \frac{\frac{1}{6}}{\frac{1}{2}} = \frac{1}{3}$$

$$P(G_3 \mid \text{host opens } 2) = \frac{\frac{1}{3}}{\frac{1}{2}} = \frac{2}{3}.$$

switching gives you a $\frac{2}{3}$ chance to win \$1000.

Expected winnings after switch

$$= \frac{2}{3} \times \$1000$$

$$\approx \$666.67$$

3.

$$(a) \quad P(A \cap B | C) = \frac{P(A \cap B \cap C)}{P(C)}$$

$$\text{also, } P(A | B \cap C) = \frac{P(A \cap B \cap C)}{P(B \cap C)}$$

$$P(B | C) = \frac{P(B \cap C)}{P(C)}$$

$$P(A | B \cap C) \cdot P(B | C) = \frac{P(A \cap B \cap C)}{P(B \cap C)} \times \frac{P(B \cap C)}{P(C)}$$

$$= \frac{P(A \cap B \cap C)}{P(C)}$$

$$= P(A \cap B | C)$$

True

$$(b) \quad P(A \cap B | C) = P(A | C) P(B | C)$$

for independent events A and B.

Solⁿ

$$P(A \cap B) = P(A) \cdot P(B)$$

$$P(A \cap B | C) = P(A | C) \cdot P(B | C)$$

False

(c) Given $P(A | D \cap B^c) > P(A | D \cap B)$ and $P(A | D^c \cap B^c) > P(A | D^c \cap B)$, $P(A | B)$ must be greater than $P(A | B^c)$.

$$\begin{aligned} P(A | B) &= \frac{P(A \cap B)}{P(B)} = \frac{P(A \cap B \cap D) + P(A \cap B \cap D^c)}{P(B)} \\ &= \frac{P(A \cap B \cap D)}{P(B \cap D)} \frac{P(B \cap D)}{P(B)} + \frac{P(A \cap B \cap D^c)}{P(B \cap D^c)} \frac{P(B \cap D^c)}{P(B)} \\ &= P(A | D \cap B) P(D | B) + P(A | D^c \cap B) P(D^c | B) \end{aligned}$$

Similarly,

$$P(A | B^c) = P(A | D \cap B^c) P(D | B^c) + P(A | D^c \cap B^c) P(D^c | B^c)$$

Now, $P(A | D \cap B^c) > P(A | D \cap B)$ $\xrightarrow{?}$ False

$$P(A | D^c \cap B^c) > P(A | D^c \cap B)$$

\therefore Not necessarily True

4.) (a) Let X take values $n = 1, 2, 3, \dots$
with probability $P_n = \frac{C}{n^3}$

Sum, $\sum_{n=1}^{\infty} \frac{1}{n^3}$ converges, so $C = \frac{1}{\zeta(3)}$

$$E[X] = \sum_{n=1}^{\infty} n P_n = C \sum_{n=1}^{\infty} \frac{1}{n^2} \quad \underline{\text{converges}}$$

$$E[X^2] = \sum_{n=1}^{\infty} n^2 P_n = C \sum_{n=1}^{\infty} \frac{1}{n} \quad \text{diverges}$$

(harmonic series)

Therefore exists.

(b) X have density fun, $f(n) = \frac{C}{n^3}$ for $n \geq 1$

$$1 = \int_1^{\infty} \frac{C}{n^3} dn = \frac{C}{2} \quad \text{so, } C = 2$$

$$E[X] = \int_1^{\infty} n \cdot \frac{2}{n^3} dn = 2$$

$$E[X^2] = \int_1^{\infty} n^2 \cdot \frac{2}{n^3} dn = \infty$$

Therefore exists $f(n) = \begin{cases} 2/n^3 & n \geq 1 \\ 0 & \end{cases}$

$$(c) \quad E(e^{-X}) \geq e^{-E[X]} = e^{-1} \approx 0.37$$

(By Jensen's inequality).

for any random variable with

$$E(X) = 1 \quad \text{we have } E(e^{-X}) \geq e^{-1} > \frac{1}{3}$$

does not exist

5. Let X_1, X_2, \dots, X_n be the prizes on n tickets drawn, each uniformly dist. on $\{1, 2, \dots, N\}$

Prob. max. of these draws is at most k :

$$P(M \leq k) = \left(\frac{k}{N}\right)^n$$

$$\begin{aligned} P(M = k) &= P(M \leq k) - P(M \leq k-1) \\ &= \left(\frac{k}{N}\right)^n - \left(\frac{k-1}{N}\right)^n \end{aligned}$$

$$E(M) = \sum_{k=1}^N k \cdot \left[\left(\frac{k}{N}\right)^n - \left(\frac{k-1}{N}\right)^n \right]$$

$$\text{Let } x = \frac{k}{N}, \text{ so } k = Nx \text{ and } dx \approx \frac{1}{N}$$

$$\begin{aligned} P(M = k) &= \left(\frac{k}{N}\right)^n - \left(\frac{k-1}{N}\right)^n \approx \frac{d}{dk} \left(\frac{k}{N}\right)^n \\ &= n \left(\frac{k}{N}\right)^{n-1} \frac{1}{N} \end{aligned}$$

$$\begin{aligned} E(M) &= \sum_{k=1}^N k \cdot n \cdot \left(\frac{k}{N}\right)^{n-1} \frac{1}{N} \\ &= nN \sum_{k=1}^N \left(\frac{k}{N}\right)^n \frac{1}{N} \end{aligned}$$

$$E(M) \approx n N \int_0^1 x^n dx = n N \left[\frac{x^{n+1}}{n+1} \right]_0^1$$

$$= n N \frac{1}{n+1} = \frac{n}{n+1} N$$

$$E(M) = \sum_{k=1}^N k \left[\left(\frac{k}{N} \right)^n - \left(\frac{k-1}{N} \right)^n \right]$$

for large N , $E(M) \approx \frac{n}{n+1} N$

6. X and Y be posⁿ of two points, dist on $[0, d]$.

$$P(|X - Y| < d/3)$$

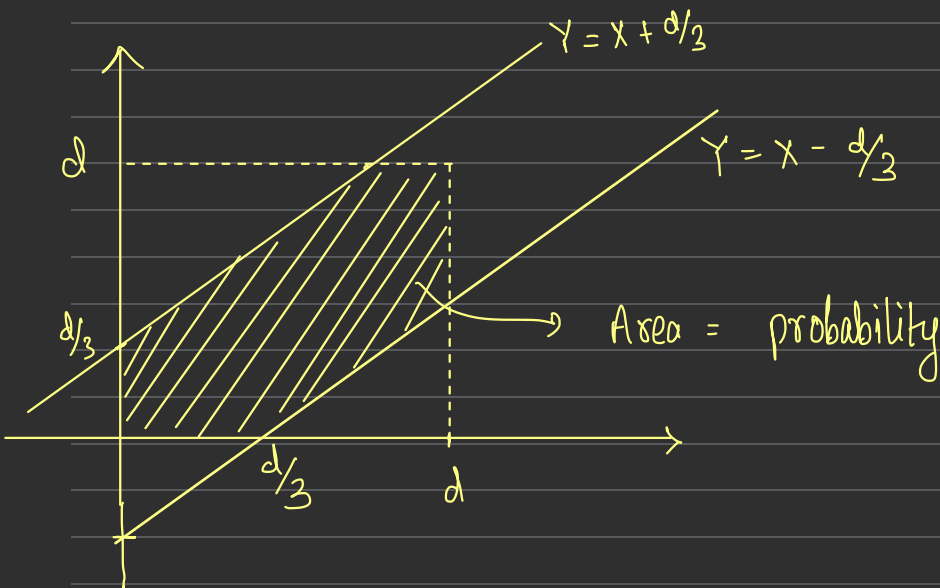
The pair (X, Y) can be repⁿ as a point on square $[0, d] \times [0, d]$

$$\text{area of square} = d^2$$

condⁿ defines region betⁿ lines

$$Y = X + d/3 \quad \& \quad Y = X - d/3$$

in square.



$$I^{st}: X \in [0, d/3]$$

$$I_1 = \int_0^{d/3} (X + d/3) dX = \int_0^{d/3} X dX + \int_0^{d/3} \frac{d}{3} dX$$

$$= \left[\frac{X^2}{2} \right]_0^{d/3} + \left[\frac{d}{3} X \right]_0^{d/3}$$

$$I_1 = \frac{d^2}{6}$$

$$I^{nd}: \left[\frac{d}{3}, d - \frac{d}{3} \right]$$

$$I_2 = \int_{d/3}^{2d/3} \frac{2d}{3} dX = \frac{2d^2}{9}$$

$$I^{rd}: \left[d - \frac{d}{3}, d \right]$$

$$I_3 = \int_{2d/3}^d (d - X) dX = \left[dX - \frac{X^2}{2} \right]_{2d/3}^d$$

$$I_3 = \frac{d^2}{6}$$

$$Area = \frac{d^2}{6} + \frac{2d^2}{9} + \frac{d^2}{6} = \frac{5d^2}{9}$$

$$Probability = \left(\frac{5d^2}{9} \right) / d^2 = \frac{5}{9}$$

7) a) At each step, rumor is passed to any one of other n people.

$$\therefore P(\text{original person is not chosen}) = 1 - \frac{1}{n}$$

$$P(\text{original person is not chosen } n \text{ times}) \\ = \left(1 - \frac{1}{n}\right)^n$$

b) At each step, rumor must be passed to a new person.

At k^{th} step : $n - k + 1$ persons left

\therefore

$$\text{probability of rumor being told } n \text{ times} \\ = \frac{n}{n} \times \frac{n-1}{n} \times \frac{n-2}{n} \dots \frac{n-(n-1)}{n}$$

$$= \prod_{k=0}^{n-1} \left(1 - \frac{k}{n}\right)$$

$$k=0$$

Now, Rumor is told to N people at each step
 \therefore Total people = nN

$$a) P(\text{original person is not chosen } nN \text{ times}) \\ = \left(1 - \frac{1}{n}\right)^{nN}$$

b) Now $nN \leq n$ \therefore every time we choose from $n-k$ people
Probability of no repetition = $\prod_{k=0}^{nN-1} \left(1 - \frac{k}{n}\right)$

$$8) P(\cap A_i^c) = P(A_1^c \cap A_2^c \cap A_3^c \dots)$$

$\therefore A_i$'s are independent

$$\begin{aligned} \therefore P(A_1^c \cap A_2^c \cap A_3^c \dots A_n^c) &= \prod_{i=1}^n P(A_i^c) \\ &= \prod_{i=1}^n (1 - P(A_i)) \end{aligned}$$

$$= 1 - P(A_1) - P(A_2) - \dots - P(A_n)$$

$$\leq e^{-\sum_{i=1}^n P(A_i)}$$

$$\therefore \boxed{1 - x \leq e^{-x}}$$

$\forall x > 0$

$$\sum P(A_i) \geq 0$$

\rightarrow on applying this inequality to each term

$$\prod_{i=1}^n (1 - P(A_i)) \leq \prod_{i=1}^n e^{-P(A_i)} = e^{-\sum P(A_i)}$$

\rightarrow proved.

g) let x & y be two independent R.V. with pdf $f_x(x)$ & $f_y(y)$

$$f_z(z) = f_x * f_y = \int_{-\infty}^{\infty} f_x(z-y) f_y(y) dy$$

$$\text{let } z = x + y$$

$$f_z(z) = \frac{d}{dz} P(x+y \leq z)$$

$$\& f_{x,y}(x,y) = f_x(x) f_y(y) \quad [\text{Independent}]$$

$$\therefore x+y = z$$

$$\begin{aligned} f_z(z) &= \int_{-\infty}^{\infty} f_{x,y}(x, z-x) dx = \int_{-\infty}^{\infty} f_x(x) f_y(z-x) dx \\ &= \int_{-\infty}^{\infty} f_x(z-y) f_y(y) dy \quad [\text{Symmetric}] \end{aligned}$$

$$\therefore f_{x+y}(z) = (f_x * f_y)(z) = \int_{-\infty}^{\infty} f_x(z-y) f_y(y) dy \quad \rightarrow \text{convolution}$$

\therefore The convolution of two distribution results in a distribution.

$$10) \int_{\Omega} \int_0^{\infty} I_{[0, x(\omega)]}(x) dx dP(\omega)$$

$$\int_0^{\infty} I_{[0, x(\omega)]}(x) dx = x(\omega)$$

$$\therefore \int_{\Omega} x(\omega) dP(\omega) = E X$$

\therefore Integrand is non-negative, we can switch Integrals

$$\therefore \int_{\Omega} \int_0^{\infty} I_{[0, x(\omega)]}(x) dx dP(\omega) = \int_0^{\infty} \int_{\Omega} I_{[0, x(\omega)]}(x) dP(\omega) dx$$

$$\int_{\Omega} I_{[0, x(\omega)]} dP(\omega) \quad \because I_{[0, x(\omega)]}(x) = \begin{cases} 1 & x < x(\omega) \\ 0 & \text{o/w} \end{cases}$$

$$\therefore \text{Inner Integral} = P(x < x(\omega)) \\ = P(x(\omega) > x) = 1 - F(x)$$

$$\therefore \int_{\Omega} \int_0^{\infty} I_{[0, x(\omega)]}(x) dx dP(\omega) = \int_0^{\infty} (1 - F(x)) dx$$

$$11) f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$(i) E e^{ux} = \int e^{ux} \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sigma\sqrt{2\pi}} dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int e^{ux - \frac{(x-\mu)^2}{2\sigma^2}} dx$$

Now,

$$\text{exponent} = ux - \frac{(x-\mu)^2}{2\sigma^2} = -\frac{1}{2\sigma^2} [(x-\mu)^2 - 2\sigma^2 ux]$$

$$= -\frac{1}{2\sigma^2} [(x-\mu)^2 - 2\sigma^2 u(x-\mu)] + u\mu$$

$$= -\frac{1}{2\sigma^2} [(x-\mu) - \sigma^2 u]^2 - \sigma^2 u^2 + u\mu$$

$$= -\frac{(x-\mu - \sigma^2 u)^2}{2\sigma^2} + \frac{\sigma^2 u^2}{2} + u\mu$$

$$\therefore E e^{ux} = \frac{1}{\sqrt{2\pi}\sigma} \int e^{-\frac{(x-\mu - \sigma^2 u)^2}{2\sigma^2}} e^{\frac{\mu u + \sigma^2 u^2}{2}} dx$$

$$= e^{\frac{\mu u + \sigma^2 u^2}{2}} \int \frac{e^{-\frac{(x-\mu - \sigma^2 u)^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma} dx$$

$$= e^{\frac{\mu u + \sigma^2 u^2}{2}} \int_{-\infty}^{\infty} \frac{e^{-\frac{(x-(\mu + \sigma^2 u))^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma} dx$$

↳ Pdf with
with $\mu + \sigma^2 u$

$$\therefore E e^{ux} = e^{\frac{\mu u + \sigma^2 u^2}{2}}$$

\therefore Integral = 1

$$(i) \quad E^x = \mu \quad \therefore \varphi(E^x) = e^{\mu}$$

$$\varphi(x) = e^{ux}$$

$$\therefore E[\varphi(x)] = E e^{ux} = e^{\mu + \frac{1}{2} u^2 \sigma^2}$$

$$\therefore \frac{1}{2} u^2 \sigma^2 \geq 0$$

$$\therefore e^{\mu + \frac{1}{2} u^2 \sigma^2} \geq e^{\mu}$$