

810

Indicator func and Fubini's Theorem



To show: $E[X] = \int_0^{\infty} [1 - F(x)] dx$

By showing: $\int_{\Omega} \int_0^{\infty} 1_{[0, X(\omega)]}(x) dx dP(\omega)$

$$= EX = \int_0^{\infty} [1 - F(x)] dx$$

$$X(\omega) = \int_0^{\infty} 1_{[0, X(\omega)]}(x) dx$$

Taking expectation value of both sides

$$E[X(\omega)] = E \left[\int_0^{\infty} 1_{[0, X(\omega)]}(x) dx \right]$$

Applying "Fubini's Theorem"

$$E[X] = \int_{\Omega} \int_0^{\infty} 1_{[0, X(\omega)]}(x) dx dP(\omega)$$

$$= \int_0^{\infty} \left(\int_{\Omega} 1_{[0, X(\omega)]}(x) dP(\omega) \right) dx$$

Now,

$$\int_{\Omega} 1_{[0, X(\omega)]}(x) dP(\omega) = P(X \geq x) = 1 - F(x)$$

$$\text{ii} \quad E(X) = \int_0^{\infty} (1 - F(x)) dx$$

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Thus,

$$E(X) = \int_0^{\infty} \int_0^{\infty} 1_{[0, x(\omega)]}(x) dx dP(x)$$

$$= \int_0^{\infty} (1 - F(x)) dx$$

87) (a) People = P_i for $i = 0, 1, 2, \dots, n$
 starting with " P_0 "

Event = "no one ever tells the rumor back to P_0 in first x steps"

At any step ' i ', there are $(n+1)$ people one of which is P_0 . So probability of " P_0 " not being chosen

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

Also,
 Every step is independent of past, future or any other step, thus according to law of independent events

$$\begin{aligned} \text{Ans: } P_x(E) &= [P(A)]^x \\ &= \left[\frac{n}{n+1} \right]^x \end{aligned}$$

(b) Probability that rumor is told k times without being repeated to any person

$P(E)$ After 1st step = 1 $\because \frac{n}{n}$

$P(E)$ after 2nd step = $\frac{n-1}{n}$

As ' $n-1$ ' are remaining

Thus after r $P(E) = \frac{n - (r-1)}{n}$
 $= \frac{n - r + 1}{n}$

$$P(E) = \left(\frac{n}{n}\right) \left(\frac{n-1}{n}\right) \left(\frac{n-2}{n}\right) \dots \left(\frac{n-r+1}{n}\right)$$

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

$$= \frac{n!}{(n-r)! \cdot n^r}$$

$$\left[\frac{n!}{(n-r)!} = (n-r)(n-r-1)(n-r-2) \dots (n-r+1) \right]$$

a) When each step tells to group of 'N' random people

$$P(E) = \frac{\binom{n}{N}}{\binom{n+1}{N}} = \frac{n(n-1) \dots (n-N+1)}{(n+1)(n) \dots (n-N+1)}$$

(b) No one is told twice

$$Pr = \prod_{k=1}^r \frac{(n+1-N(k-1)-1)}{(n+1) \binom{n}{N}}$$

Q4/a) A Discrete random variable x for which $E(x)$ is finite but $E(x^2) \rightarrow \infty$ (b)

Sol Let x be random variable such that we define $x \in N$

$$N = \{1, 2, 3, \dots\} \quad [\text{finite}]$$

$$P(x=n) = \frac{c}{n^p} \quad [n \geq 1]$$

~~Now,~~

$$E(x) = \sum_{n=1}^{\infty} n \cdot \frac{c}{n^p} = c \sum_{n=1}^{\infty} \frac{1}{n^{p-1}}$$

$$E(x^2) = \sum_{n=1}^{\infty} \frac{n^2 c}{n^p} = c \sum_{n=1}^{\infty} \frac{1}{n^{p-2}} \quad (c)$$

Let $p > 2$

Therefore $\sum \frac{1}{n^{p-1}}$ will converge

But

$\sum \frac{1}{n^{p-2}}$ will diverge

Ex] say $p = 3$ (say)

$$c \sum_{n=1}^{\infty} \frac{1}{n} \rightarrow \infty \quad \text{but} \quad c \sum_{n=1}^{\infty} \frac{1}{n^2} \text{ will}$$

be finite.

(b) $f(x) = \frac{c}{x^p} \quad x > 1 \text{ and } p > 1$

$$E(x) = \int_1^{\infty} x \cdot \frac{c}{x^p} dx = c \int_1^{\infty} x^{(1-p)} dx$$

$$E(x) = \int_1^{\infty} x^{2-p} \frac{c}{x^p} dx = c \int_1^{\infty} x^{2-p} dx$$

Now Let $p \neq 2$ say $p = 2.5$

Then,
$$\begin{cases} E(x) < \infty \\ E(x^2) = \infty \end{cases}$$
 Thus, exist

(c) $E(e^{-x}) > e^{-E(x)} > e^{-1} > 0.3679$

Thus for Random variable with $E(x) \geq 1$ we must have

$$E(e^{-x}) > \frac{1}{3}$$

Ans: Impossible one does not exist

Q3] (a) $P(A \cap B | C) = P(A | B \cap C) P(B | C)$

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

multiply by $P(B | C)$

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

Hence,
True

$$\times \frac{P(B \cap C)}{P(C)}$$

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

A & B are independent events

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

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

Not always true, independent events A and B does not guarantee independence of C

Hence, True

Q3] (c) A is more likely to occur when B is absent

i.e. B^c ~~for~~

for D $A|B^c > A|B$

for B^c $A|B^c > A|B$

This probability is just ~~and~~ weighted sum of regional probabilities

Ans: Hence $P(A|B^c) > P(A|B)$

Hence, True

Q8]

Independent events have independent components

$$P\left(\bigcap_{i=1}^n A_i^c\right) = \prod_{i=1}^n P(A_i^c)$$

so,

$$\prod_{i=1}^n P(A_i^c) = \prod_{i=1}^n [1 - P(A_i)]$$

for any real no. $1 - x \leq e^{-x}$

This gives

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

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

Thus,

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