

EXPECTATION

HW. find out. binomial distribution geometric distribution

prove that. $E(ax+b) = aE(x) + b$

$$Y = aX + b$$

$$Y(\omega) = aX(\omega) + b$$

RECAP

- probability axioms
- derived few important results
- conditional probability
- random variables (discrete & continuous)
- pmf, pdf, cdf

$$(i) \quad 0 \leq p(E) \leq 1$$

$$(ii) \quad p(\Omega) = 1$$

$$(iii) \quad \text{if } E_1 \cap E_2 = \emptyset \quad E_1, E_2 \in \mathcal{F}$$

$$p(E_1 \cup E_2) = p(E_1) + p(E_2)$$

$$p(A|B) = \frac{p(A \cap B)}{p(B)} \quad \text{if } p(B) > 0$$

gaussian distribution

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad x \in (-\infty, \infty)$$

where, μ = mean, σ = Standard deviation.

$\mu=0$ $\sigma=1$ is called normal distribution

EXPECTATION OF RV

X

e.g. i toss a coin n times, how many heads we expect?
 $N/2$.

$$\Omega = \{ \underbrace{HHH \dots H}_n, \underbrace{HTH \dots H}_n, \dots, \underbrace{TTT \dots T}_n \} \quad 2^n \text{ outcomes}$$

$$X: \Omega \rightarrow \mathbb{R} \quad X(\omega) = \{0, 1, 2, \dots, n\}$$

$$X(\omega) = \# \text{ heads in } \omega \text{ of } \Omega$$

↪ sample

K experiments

1 = experiment is tossing coin n times

k_1 times 0 H $X(x)$ $\omega_1 \in \Omega$

k_2 times 1 H ω_2

⋮

⋮

$$k_1 + k_2 + \dots + k_n = K$$

k_{n+1} times n H

ω_{n+1}

$$E(X) = \frac{0 \cdot K_1 + 1 \cdot K_2 + \dots + n \cdot K_{n+1}}{K}$$

sample avg.

$$\lim_{K \rightarrow \infty} = 0 \cdot \lim_{K \rightarrow \infty} \frac{K_1}{K} + 1 \cdot \lim_{K \rightarrow \infty} \frac{K_2}{K} + \dots + n \cdot \lim_{K \rightarrow \infty} \frac{K_{n+1}}{K}$$

$$\lim_{K \rightarrow \infty} \frac{K_1}{K} = p(X=0)$$

$$\lim_{K \rightarrow \infty} \frac{K_i}{K} = p(X=i-1) \quad i = \{1, \dots, n+1\}$$

$$E_X = \sum_{x \in X(\omega)} x p(X=x) \quad \text{expectation for DRV.}$$

$$X(\omega) = (x_1, \dots, x_n)$$

$$E_X = \int_{x: p(x) > 0} x \cdot f_x(x) dx$$

expectation of CRV

$$p: x \rightarrow x+dx$$

$$= f_x(x) dx$$

HW find out binomial distribution
geometric distribution.

prove that $E(aX+b) = aE_X + b$

$$Y = aX + b$$

$$Y(\omega) = aX(\omega) + b$$

$$\begin{aligned} p(Y=y) &= p(\{\omega \mid Y(\omega) = y\}) \\ &= p(\{\omega \mid aX(\omega) + b = y\}) \end{aligned}$$

bernoulli RVeach toss $p(H) = p$

$$X = 1 \quad p$$

if H, $X = 1$

$$= 0 \quad \underbrace{1-p}_{\text{prob.}}$$

if T, $X = 0$

$$E = 1 \cdot p + 0 \cdot (1-p) = p$$

if H $Y = 5 \cdot 1 + 3 = 8$

$$Y = 5x + 3$$

if T $Y = 5 \cdot 0 + 3 = 3$

$$EY = 8p + 3(1-p)$$

E is linear operator,

$$= 5p + 3$$

binomial RV (Ch.p.) $X = \# \text{ successes in } n \text{ independent bernoulli trials.}$ $X_i = 1$ if i th bernoulli trial is success,
 $= 0$ otherwise.

$$X = \sum_i X_i$$

$$EY = \sum_i EX_i = \sum_i p = n \cdot p$$

HW

$$E(g(X)) = \sum_x g(x) p(X=x)$$

Poisson RV

$$X \in \{0, 1, 2, \dots, \infty\}$$

$$p(X=k) = \frac{e^{-\lambda} \lambda^k}{k!}$$

$$\sum_k p(X=k) = 1$$

$$\underline{EX} = \sum_{k=0}^{\infty} k \cdot \frac{e^{-\lambda} \lambda^k}{k!} = \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{(k-1)!}$$

$$= \sum_{k'=0}^{\infty} \frac{e^{-\lambda} \lambda^{k'+1}}{k'!}$$

$$= \lambda e^{-\lambda} \underbrace{\sum_{k'=0}^{\infty} \frac{\lambda^{k'}}{k'!}}_{e^{\lambda}} = \underline{\underline{\lambda}}$$

VARIANCE

how much spread the distribution values are.

$$E(X - EX)^2 \xrightarrow{\text{prove}} EX^2 - [EX]^2$$

M_1 - first moment

mean (μ) is called the first moment of RV.

EX^2 is called as the second moment.

moment generating functions

EX^i is called the i th moment.

$$Ee^{tx} = 1 + tM_1 + \frac{t^2}{2!} M_2 + \dots$$

MEMORYLESS RV

exponential RV $f_x(x) = \lambda e^{-\lambda x}$ if $x \geq 0$

$= 0$

O.W.

$$\text{cdf } F_x(x) = 1 - e^{-\lambda x} \quad x \geq 0$$

$$F_x(x) = P(X \leq x)$$

$$P(X > t | X > t_0)$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \frac{P(X > t)}{P(X > t_0)}$$

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

$$P(A^c) = 1 - P(A) \quad = \frac{e^{-\lambda t}}{e^{-\lambda t_0}}$$

$$P(X > t - t_0)$$

$$= \frac{1 - (1 - e^{-\lambda t})}{1 - (1 - e^{-\lambda t_0})}$$

$$= e^{\lambda(t_0 - t)}$$

$$= e^{-\lambda(t - t_0)}$$

$$\therefore P(X > t | X > t_0) = P(X > t - t_0)$$

markov property

game play learning.

physics engine

classmate

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X : time at which 1st packet arrives

$$t_0 = 100 \quad t = 105$$

$$t_0' = 5 \text{ sec}$$

$$t' = 10 \text{ sec.}$$

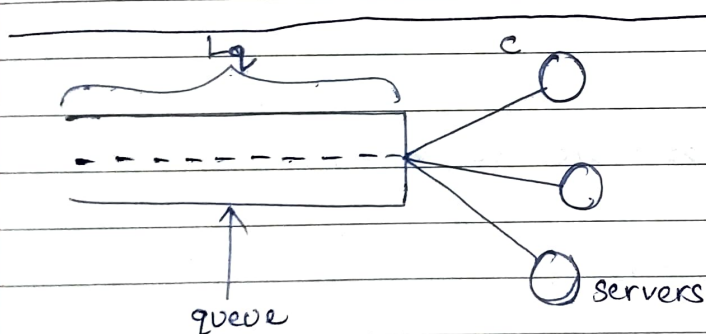
doesn't matter,

$$p(X > t \mid X > t_0) = p(X > t_0' \mid X > t_0)$$

memory less.

$$p(X_{t_{n+1}} = X_{t_n} = X_{t_{n-1}} \dots X_{t_0} = x_{t_0})$$

$$= p(X_{t_{n+1}} = x_{t_n} \mid X_{t_n} = x_{t_n}) \leftarrow \text{markov}$$



L_s : average or expected packet / customers / requests at servers.

L_q : —————
in the queue.

time spent
in the system,

L = avg. or expected in the system.

$$L = L_q + L_s$$

$$w = w_s + w_q$$

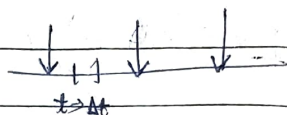
average rate of arrival λ

$$\lim_{\Delta t \rightarrow 0} \frac{o(\Delta t)}{\Delta t} \rightarrow 0$$

(i) from $t \rightarrow t + \Delta t$

$$p(\text{one arrival}) = \lambda \Delta t + o(\Delta t)$$

(ii) $p(\text{more than one arrival})$



(iii) # arrivals in non-overlapping intervals are independent of each other



state n : till time t , n packets arrived.

$q_k(t)$ = prob. that there are k arrivals till time t .

$$k \geq 1 \quad q_k(t + \Delta t) = q_{k-1}(t) \times \lambda \Delta t + q_k(t) (1 - \lambda \Delta t)$$

$$q_k(t + \Delta t) - q_k(t) = \lambda \Delta t (q_{k-1} - q_k)$$

$$\frac{q_k(t + \Delta t) - q_k(t)}{\Delta t} = \lambda q_{k-1} - \lambda q_k \quad \text{--- (1)}$$

for $k=0$.

$$q_0(t + \Delta t) = q_0 (1 - \lambda \Delta t)$$

$$\frac{q_0(t+\Delta t) - q_0(t)}{\Delta t} = -\lambda q_0(t) \quad \text{--- (2)}$$

$$\lim_{\Delta t \rightarrow 0} \text{ in (2) } \frac{dq_0(t)}{dt} = -\lambda q_0(t)$$

$$q_0(t) = e^{-\lambda t} \times c$$

$$q_0(0) = 1 \Rightarrow c = 1$$

$$q_0(t) = e^{-\lambda t}$$

$$\lim_{\Delta t \rightarrow 0} \text{ in eq. (1)}$$

$$\frac{dq_k(t)}{dt} = \lambda q_{k-1}(t) - \lambda q_k(t)$$

for $k=1$

$$\frac{dq_1(t)}{dt} = \lambda q_0(t) - \lambda q_1(t)$$

$$\frac{dq_1(t)}{dt} = -\lambda q_1(t) + \lambda e^{-\lambda t}$$

$$q_1(t) = (\lambda t) e^{-\lambda t}$$

\vdots

$$q_k(t) = \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$

poisson distribution.

(memory less)