

# Chapter 6.

1. (a)  $P_{R(r)} = \binom{n}{r} p^r (1-p)^{n-r}$

$$E(R) = \sum_{r=0}^n r \cdot P_{R(r)} = E(\sum R_i) = np$$

$$\text{Var}(R) = np(1-p)$$

(b)  $P(B) = P(\text{first item on Red car, the others on Green car}) + P(\text{first Green, others Red})$   
 $= p \cdot (1-p)^{n-1} + (1-p) \cdot p^{n-1}.$

(c)  $P(c) = P(\text{1 Red, } n-1 \text{ Green}) + P(\text{1 Green, } n-1 \text{ Red})$   
 $= \binom{n}{1} p \cdot (1-p)^{n-1} + \binom{n}{1} p^{n-1} \cdot (1-p) \quad (\text{if } n \neq 2)$

$$\text{if } n=2, \quad P(c) = \binom{2}{1} p \cdot (1-p)^{n-1} = 2p(1-p).$$

(d)  $R-G = R-(n-R) = 2R-n.$

$$E(R-G) = 2E(R)-n = 2np-n,$$

$$\text{Var}(R-G) = 4\text{Var}(R) = 4np(1-p).$$

(e) assume that  $R = X_1 + X_2 + \dots + X_n$ ,

$X_1, \dots, X_n$  are independent.

$$P(X_3 + \dots + X_n \mid X_1, X_2 = 1) = P(X_3 + \dots + X_n)$$

$$E(R \mid X_1, X_2 = 1) = E(2 + X_3 + \dots + X_n) = 2 + (n-2)p.$$

$$\text{Var}(R \mid X_1, X_2 = 1) = (n-2)p \cdot (1-p).$$

$$2. (a) P = \binom{6}{2} \cdot \left(\frac{3}{4}\right)^4 \cdot \left(\frac{1}{4}\right)^2$$

(b)  $T_3$  denotes the number of all quizzes that contains failing up to 3 times. the last situation is 3rd failing.

~~xi denotes~~

$$x_i \triangleq T_i - T_{i-1}, T_0 = 0, p \triangleq \frac{1}{4}.$$

so  $Y \sim \text{Pascal}$ .

$$P_{T_3}(y) = \binom{y-1}{2} p^2 \cdot (1-p)^{y-3} \cdot p$$

$$= \binom{y-1}{2} p^3 \cdot (1-p)^{y-3}.$$

$$E(T_3) = E(x_1 + x_2 + x_3) = \frac{3}{p} = 12.$$

E(number of quizzes that he will pass) =  $12 - 3 = 9$ .

(c)  ~~$P(A) = P(\text{he failed in first 7 quizzes } 1 \text{ time}) = \binom{7}{1} p \cdot (1-p)^6$~~

~~$P(B) = P(\text{he fails in first 7 quizzes } 2 \text{ times}) = \binom{7}{2} p^2 \cdot (1-p)^5$~~

~~$P(2^{\text{nd}} \text{ times}) = P(2^{\text{nd}} \text{ failing in } 8^{\text{th}} \text{ quiz}) + P(2^{\text{nd}} \text{ not failing, in } 9^{\text{th}} \text{ quiz})$~~

$$= P(A) \cdot p + P(A) \cdot (1-p) \cdot p + (1-P(A)) \cdot p^2$$

$$= p^2 + 2(p-p^2) \cdot P(A)$$

$P(A) \triangleq P(\text{he failed 1 times in the first 7 quizzes})$

$$P(C \text{ problems}) = P(A) \cdot p^2 = \binom{7}{1} p^3 \cdot (1-p)^6.$$

$$= \binom{7}{1} \left(\frac{1}{4}\right)^3 \cdot \left(\frac{3}{4}\right)^6.$$

(d)  
①

$P(\text{he pass 2 times in a row})$

$$= 1 - P(\text{he never pass 2 times in a row})$$

set  $H_k$  be the probability of never passing 2 times in a row.

$S \checkmark$ : Pass  $\bar{F}$ : Fail. We let  $(SF)$  must together

$$P(H_1) = P(\bar{F}) = \frac{1}{4}$$

$$P(H_2) = P(H_1) \cdot P(\bar{F}) + P(SF) = \frac{1}{4}$$

$$\begin{aligned} P(H_n) &= P(H_{n-1}) \cdot P(\bar{F}) + P(H_{n-2}) \cdot P(SF) \\ &= \frac{1}{4}P(H_{n-1}) + \frac{3}{16}P(H_{n-2}) \end{aligned}$$

~~since~~ since  $H_n$  为 压缩映射 (compressed mapping)

$$P(H_n) = \frac{16}{9}^{\text{or } 0}, \text{ as } n \rightarrow \infty.$$

$$\Rightarrow \boxed{P(H)} \text{ written as } P(H) = \lim_{n \rightarrow \infty} P(H_n) = \frac{16}{9}^{\text{or } 0}.$$

we can devide  $H_\infty$  into infinite parts of  $H_k$ ,  $k$  is a constant

$$P(H_\infty) \leq (P(H_k))^n \quad (n \rightarrow \infty).$$

we have  $P(\text{pass 2 times in a row}) = 1$ .

②  $P(2 \text{ fails in 2 row before 2 passing in a row})$

$= P(\text{FF first time, and pass 2 times in a row})$

$$= \boxed{P(H)} \sum_{n=2}^{+\infty} P(\text{SF SF -- FF or FS FS -- FF in } n \text{ quizzes})$$

$$= \left(\frac{1}{4}\right)^2 \left(1 + \frac{3}{4} + \frac{3}{4} \cdot \frac{1}{4} + \left(\frac{3}{4}\right)^2 \cdot \frac{1}{4} + \left(\frac{3}{4}\right)^2 \cdot \left(\frac{1}{4}\right)^2 + \dots\right) = \left(\frac{1}{4}\right)^2 \cdot \frac{\frac{7}{4} \cdot (1)}{1 - \frac{3}{16}} = \frac{7}{52}. \square$$

3.

$$(a) P(1, B) = P_B \cdot P_{1|B} = \frac{5}{6} \cdot \frac{2}{5} = \frac{1}{3}.$$

$$P(A) = (1 - P(1, B))^3 \cdot P(1, B)$$

$$= \left(\frac{2}{3}\right)^3 \cdot \frac{1}{3}$$

$$(b) P(B \text{ prob.}) = \frac{5}{6} \cdot \frac{1}{6}$$

$$(c) \quad \cancel{P(1 \text{ task})} \quad P(1 \text{'s task } \cancel{\text{is}} \mid n^{\text{th}} \text{ task}) = \frac{1}{3}.$$

$$P \triangleq \frac{1}{3},$$

$T_k$ : interarrival  $k^{\text{th}}$

$$Y_k \stackrel{A}{=} T_1 + \dots + T_k.$$

$$E(T_1 + \dots + T_5) = 5, E(T_5) = 15. \quad | \text{ if } X \sim \text{Geometric}(p), E(X) = \frac{1}{p}.$$

$$(d) \quad \cancel{P(\text{busy slots} \mid \cancel{\text{user}} \leq 5)} \quad |$$

$E(\text{user } 1 \leq 5)$  about Busy slots.

$\cancel{Y_i}$  denotes number of busy slots when  $i^{\text{th}}$  1's task starts

$$T_i = Y_i - Y_{i-1}, \quad Y_0 = 0.$$

$$E(\text{user } 1 \leq 5) = E(Y_5) = 5 \cdot E(T) = 5 \cdot \frac{5}{2} = \frac{25}{2}.$$

(e)  $\cancel{Y_i \text{ denotes}}$

$Y_i$  denotes busy number of tasks until  $i^{\text{th}}$  task from user 1

$Z$  denotes number of tasks from user 2 until 5th 1's task.

$$Z = Y_5 - 5. \quad P_{Y_5}(y) = \binom{y-1}{4} \cdot \left(\frac{2}{5}\right)^4 \cdot \left(\frac{3}{5}\right)^{y-5} \cdot \left(\frac{2}{5}\right)$$

$$P_Z(z) = \binom{z+4}{4} \cdot \left(\frac{2}{5}\right)^4 \cdot \left(\frac{3}{5}\right)^z$$

$$E(Y_5) = 5 E(T), \quad \text{Var}(Y_5) = 5 E(T).$$

$$E(Z) = 5E(T) - 5 = \frac{25}{2} - 5 = \frac{15}{2}.$$

$$\begin{aligned}\text{Var}(Z) &= 5\text{Var}(T) = 5 \cdot \frac{1-(2/5)}{(2/5)^2} \\ &= \frac{75}{4}. \quad \square\end{aligned}$$

4. (a) let  $Y_k$  be the number of trials up to  $k$ th success.

$$T_k = Y_k - T_{k-1}, \quad Y_0 = 0$$

let  $X_k$  be the number of ~~failures~~ failures -- --

$$X_k = Y_k - k.$$

$$\underline{P_{T_k}(y)} = \binom{y-1}{k-1} p^k (1-p)^{y-k}$$

$$P_{T_k}(y) = \binom{y-1}{k-1} p^k \cdot (1-p)^{y-k}.$$

$$P_{X_k}(x) = P_{T_k}(x+k) = \binom{x+k-1}{k-1} p^k \cdot (1-p)^x$$

$$(b) \quad \mathbb{E}(X_k) = k \cdot E(T) - k$$

$$= k \cdot \frac{1}{p} - k.$$

$$\text{Var}(X_k) = k \cdot \frac{1-p}{p^2}.$$

(c)  $P(i\text{th failure before } r\text{th success})$

$$\begin{aligned}&= \sum_{x=1}^{+\infty} P_{X_k}(x) = 1 - \sum_{x=0}^{i-1} P_{X_k}(x) \\ &= 1 - \sum_{x=0}^{i-1} \left( \binom{x+k-1}{k-1} \cdot p^k \cdot (1-p)^x \right). \quad \square\end{aligned}$$

5. Let  $t$  be the time I am in the room.

$B$  be the latest time he won the game.

$F$  be the future time he win the game  
first

since time  $\leq t$  is independent with time  $> t$ .

$P_{B-t} \sim \text{Geometric}(p)$ .

$$\cancel{P(B-t)} \quad P_{B-t}(y) = (1-p)^{y-1} \cdot p$$

$$\text{symmetrically, } \cancel{P_{t-B}(x)} = P_{t-B}(x) = (1-p)^{x-1} \cdot p$$

$$\cancel{P_{F-B}(z)} = P((F-t)+(t-B)-z)$$

$$z = F - B = (F - t) + (t - B),$$

$Z \sim \text{Pascal}(p, z)$ . since  $\text{Bernoulli}(p) + \text{Bernoulli}(p) \sim \text{Pascal}(p, 2)$

$$P_Z(z) = \binom{z}{1} \cdot p^2 \cdot (1-p)^{z-1}$$

$$= z \cdot p^2 \cdot (1-p)^{z-1}. \quad \square$$

6.  $Y = X_1 + X_2 + X_3 + \dots + X_N$ .

$X_i \sim \text{Geometric}(p)$ ,  $N \sim \text{Geometric}(q)$ .

$\bullet$   $X_1, \dots, X_N, N$  are independent. independent.  
independent.

$$\textcircled{1} \quad P_Y(y) = \sum_{N=1}^{+\infty} P_{T,N}(y, n) = \sum_{n=1}^{+\infty} P_{T,N}(y|n) \cdot P_N(n)$$

$$= \cancel{\sum_{n=1}^{+\infty} \binom{y-1}{n-1}} \cdot p^n \cdot (1-p)^{y-n} \cdot \cancel{(1-q)^{n-1}} \cdot q$$

$$Pr(Y) = \sum_{n=1}^{\infty} \binom{y-1}{n-1} \cdot \left(\frac{P(1-q)}{1-p}\right)^n \cdot (1-p)^y \cdot \frac{q}{1-q}.$$

$$= \sum_{n=0}^{\infty} \binom{y-1}{n} \cdot \left(\frac{P(1-q)}{1-p}\right)^{n+1} \cdot (1-p)^y \cdot \frac{q}{1-q}$$

$$= \left(\frac{P(1-q)}{1-p} + 1\right)^{y-1} \cdot \left(\frac{P(1-q)}{1-p}\right) \cdot (1-p)^y \cdot \frac{q}{1-q}$$

$$= (1-pq)^{y-1} \cdot pq$$

$\Rightarrow Y \sim \text{Geometric}(pq)$ .  $\square$

② Another solution:

(Another solution by hints.)

(Another solution by hints).

~~We can split each arrival time~~

~~we can assume that  $x_i$  denotes the  $i^{th}$  interarrival times.~~

~~and it fails  $(x_i - 1)$  times with probability  $(1-p)$ , and success with  $p$ .~~

$N$  denotes the  $n^{th}$  ~~success~~ is accepted with probability  $q$ , and each arrival can be accept with  $pq$ . prob..

Hence we get  $Y$  is geometric distribution with parameter  $pq$ .  $\square$

$$P(Y=y) = \sum_{n=1}^y \binom{y-1}{n-1} \cdot \left(\frac{p(1-q)}{1-p}\right)^n \cdot (1-p)^y \cdot \frac{q}{1-q}.$$

$$= \sum_{n=0}^{y-1} \binom{y-1}{n} \cdot \left(\frac{p(1-q)}{1-p}\right)^{n+1} \cdot (1-p)^y \cdot \frac{q}{1-q}$$

$$= \left(\frac{p(1-q)}{1-p} + 1\right)^{y-1} \cdot \left(\frac{p(1-q)}{1-p}\right) \cdot (1-p)^y \cdot \frac{q}{1-q}$$

$$= (1-pq)^{y-1} \cdot pq$$

~~⇒~~  $Y \sim \text{Geometric}(pq)$ .  $\square$

② Another solution:

(Another solution by hints.)

(Another solution by hints).

~~We can split ~~the~~ arrival time~~

~~we can assume that  $X_i$  denotes the  $i^{\text{th}}$  interarrival times.~~

~~and it fails  $(X_i-1)$  times with probability  $(1-p)$ , and success with  $p$ .~~

$N$  denotes the  $n^{\text{th}}$  ~~success~~ is accepted with probability  $q$ , and each arrival can be accept with  $pq$ . prob..

Hence we get  $Y$  is geometric distribution with parameter  $pq$ .  $\square$

7. (a) ~~If we can prove~~

~~if~~ ~~if~~

~~since~~  $P(0 \leq T < \frac{1}{2}) = P(\frac{1}{2} \leq Y < 1) = \frac{1}{2}$   
we have

& since  $P(0 \leq T < \frac{1}{2}) = P(Y=0, X_2=0, \dots)$

$$P(\frac{1}{2} \leq Y < 1) = P(Y=1 | T=0, X_2=0, \dots)$$

and we can also get  $P(0 \leq T < \frac{1}{4}) = P(\frac{1}{4} \leq T < \frac{1}{2}) = \dots = \frac{1}{4}$

we have  $P(\frac{1}{2^k} \leq T < \frac{i+1}{2^k}) = \frac{1}{2^k}$ , for  $\forall k \geq 1, 0 \leq i < 2^k$ .

and we can conclude  $P(T < \frac{1}{2^k}) = \frac{1}{2^k}$ .

for  $\forall y \in (0, 1)$ , we have a series  $\{a_n\}_{n=1}^{\infty}$ ,

$$a_n = \frac{b_n}{2^n}, \quad b_n \in \mathbb{N}^+, \quad \text{and } a_n \leq y < a_n + \frac{1}{2^n},$$

we have ~~if~~  $P(Y \leq a_n) \leq P(Y \leq y) \leq P(Y \leq a_n + \frac{1}{2^n})$

$$\Rightarrow a_n \leq P(Y \leq y) < a_n + \frac{1}{2^n}$$

$$\Rightarrow \lim_{n \rightarrow \infty} a_n \leq P(Y \leq y) \leq \lim_{n \rightarrow \infty} a_n + \frac{1}{2^n} = y$$

$$\Rightarrow P(Y \leq y) = y, \quad Y \sim \text{uniform distribution in } (0, 1).$$

(b)  $P(T < \frac{1}{2}) = P(Y=0 | \dots) = P(X_1=0) = \frac{1}{2}.$

$$P(Y \geq \frac{1}{2}) = P(X_1=1) = \frac{1}{2}.$$

$$\Rightarrow X_1 \sim \text{Uniform}\{0, 1\}.$$

for any  $n \geq 2$ , we have  ~~$P(Y \leq \sum_{k=1}^{n-1} 2^{-k} X_k)$~~

$$\cancel{P\left(\sum_{k=1}^{n-1} 2^{-k} \cdot x_k \leq Y < \sum_{k=1}^{n-1}\right)}$$

$$\cancel{P\left(\sum_{k=1}^{n-1} 2^{-k} \cdot x_k + 2^{-n} \cdot x_n \leq Y < \sum_{k=1}^{n-1} 2^{-k} \cdot x_k + 2^{-n} \cdot (x_n+1)\right)}$$

$$\cancel{\sum_{k=1}^{n-1} 2^{-k} \cdot x_k \leq Y < \sum_{k=1}^{n-1} 2^{-k} \cdot x_k + 2^{-(n-1)}}$$

$$= \cancel{\sum} = P(0, x_1, \dots, x_{n-1}, x_n = \dots | 0, x_1, \dots, x_{n-1} = \dots)$$

$$= P(x_1 = x_1, x_2 = x_2, \dots, x_{n-1} = x_{n-1}, x_n = x_n = \dots | x_1 = x_1, \dots, x_n = x_n, \cancel{x_n = x_n})$$

$$\cancel{= P(x_n = x_n) = P(x_n = 0) = P(x_n = 1)}$$

since  $x_1, x_2, \dots, x_n, \dots$  are independent.  
are independent.

$$\Rightarrow \cancel{P(x_n = 0) = P(x_n = 1) = \frac{1}{2}}$$

since  $\cancel{P\left(\sum_{k=1}^{n-1}\right)}$  assume that we have

$$P(x_1 = x_1, \dots, x_i = x_i, \dots) = \frac{1}{2^i}, \quad \forall i \leq n-1$$

$$P(x_i = x_i) = \frac{1}{2}, \quad \forall i \leq n-1.$$

$$\text{And } P\left(\sum_{k=1}^{n-1} 2^{-k} x_k \leq Y < \sum_{k=1}^{n-1} 2^{-k} \cdot x_k + 2^{-n}\right) = \frac{1}{2^n}.$$

$$= P(x_1 = x_1, \dots, x_n = x_n)$$

$$P(x_n = x_n) = \sum_{x_1} \sum_{x_2} \dots \sum_{x_{n-1}} P(x_1 = x_1, \dots, x_n = x_n) \\ = \frac{1}{2},$$

$$\Rightarrow P(x_n = 0) = P(x_n = 1) = \frac{1}{2}.$$

through inductive method, we have  $P(x_n = 0) = P(x_n = 1) = \frac{1}{2}$ ,

$x_n \sim \text{Bernoulli}\{0, 1\}$ , for  $\forall n \in \mathbb{N}$ .  $\square$

8. During rush hour, from 8. a.m. to 9. a.m. traffic accidents occur according to a Poisson process with a rate of 5 accidents per hour.

8. - 9. a.m. : Poisson(5)

9 - 10 a.m. : Poisson(3)

10 - 11 a.m. : Poisson(3)

$$\begin{aligned} M_{X(s)} &= E(e^{sX}) = \sum_{x=0}^{+\infty} e^{sx} \cdot e^{-\lambda} \cdot \frac{\lambda^x}{x!} \\ &= \sum_{x=0}^{+\infty} e^{-\lambda} \cdot \frac{(e^s \cdot \lambda)^x}{x!} \\ &= e^{-\lambda} \cdot e^{e^s \cdot \lambda} = e^{(e^s - 1)\lambda} \end{aligned}$$

$$M_{X_1(s)} \cdot M_{X_2(s)} = e^{(e^s - 1)\lambda_1} \cdot e^{(e^s - 1)\lambda_2} = e^{(e^s - 1)(\lambda_1 + \lambda_2)}$$

$$\Rightarrow \text{Poisson}(\lambda_1) + \text{Poisson}(\lambda_2) = \text{Poisson}(\lambda_1 + \lambda_2)$$

$$\text{Poisson}(5) + \text{Poisson}(3) + \text{Poisson}(3) = \text{Poisson}(11). \quad \square$$

9. time ~ Exponential

5 tennis courts ?

(4 + 8k)?

If we can assume that each table of 5 tables are busy all the time, so we get  $E(1 \text{ pairs}, 1 \text{ table}) = \frac{5}{5+5} = \frac{1}{2}$

?

•  $\square$

(10.

$$(a) P(0 \text{ fish} | \text{an hour}) = e^{-\lambda}$$

$$P(\text{stay} \geq 2 \text{ hours}) = P(0 \text{ fish} | 1 \text{ hour})^2 = e^{-2\lambda} = e^{-1.2}$$

$$\begin{aligned} (b) P(2-5 \text{ hours}) &= 1 - P(T \leq 5 \text{ hours}) = P(T < 5 \text{ hours}) \\ &= P(T \geq 2 \text{ hours}) - P(T \geq 5 \text{ hours}) \\ &= e^{-1.2} - e^{-3}. \end{aligned}$$

(c) he catches at least 2 fish in the first 2 hours

$$\begin{aligned} \sum_{k=2}^{+\infty} P(F=k) &= 1 - P(F=0) - P(F=1) \\ &= 1 - e^{-2\lambda} - e^{-2\lambda}(2\lambda) \end{aligned}$$

(d)  $\tilde{E}(X) = E(*E(X|F))$ .  $F$  denotes the

$$E(X) = E(X | 0 \text{ fish in the first 2 hours}) \cdot P(0 \text{ fish})$$

$$+ E(X | \geq 1 \text{ fish, 2 hours}) \cdot P(\geq 1 \text{ fish})$$

$$= 1 \cdot P(0 \text{ fish}) + P(\geq 1 \text{ fish}) \cdot E(\text{Poisson}(\lambda))$$

$$= e^{-2\lambda} + (1-e^{-2\lambda})(2\lambda)$$

$$= 2\lambda + (1-2\lambda) \cdot e^{-2\lambda}$$

$$= 1.2 - 0.2 \cdot e^{-1.2}$$

Time  $\sim \text{Poisson}(k, t)$

(e) Due to the memoryless property of Poisson distribution

$$P(\text{Time} \leq t) = 1 - P(0, t) = 1 - e^{-\lambda t}, f_T(t) = \lambda \cdot e^{-\lambda t}, E(T) = \frac{1}{\lambda},$$

$$E(T+4) = 4 + \frac{1}{\lambda}$$

1. ~~depart 出发~~ (depart 离开).  
~~(depart 出发)~~

when a customer departs from the bookstore, he will buy a book with probability  $p$ .

$P(k, y)$  denotes  $k$  customers depart in  $y$  time.

~~$P(k, y)$~~  is the probability of selling books in  $y$  time obeys Poisson( $\lambda p$ )

(a)  $P(T \leq t) \text{ first customer} = 1 - e^{-\lambda t}$ .

$$P(T \leq t) \text{ first book} = 1 - e^{-\lambda p t}.$$

b)  $e^{-\lambda p} \quad 1 - P(T \leq 1)$

(c)  $E(\text{Poisson}(\lambda p)) = E(\text{Poisson}(\lambda p)) = \lambda p. \quad \square$   
 $\text{Poisson}(\lambda p).$

+2. ~~Go! Go! Poisson( $\lambda p$ )~~

12. ~~Go!~~ A pizza serves  $n$  different types of pizza, and is visited by a number  $k$  of customers in a given period of time, where  $k$  is a Poisson random variables with mean  $\lambda$ .

$k$  customers :  $k \sim \text{Poisson}(\lambda)$ .

$n$  types of pizza. : uniform distribution.

~~we assume that  $x_k$  denotes the  $k$ th different pizza.~~

~~the number of customers choose~~

~~$T_k = x_k - x_{k+1}, T_1 = 1, T_k \sim \text{Geometric}\left(\frac{n-k+1}{n}\right)$~~

$$P(T_k = t) = \frac{n-k+1}{n} \cdot \left(\frac{k-1}{n}\right)^{t-1},$$

$$\bar{T}_k = T_1 + T_2 + \dots + T_k$$

$$P(T_k = k) = P(\bar{T}_k \leq k < \bar{T}_{k+1})$$

$$E(k) = \sum_{k=1}^n k P$$

assume that  $X_k (1 \leq k \leq n)$  denotes the number of customers who choose  $k$ th pizza.

$$X_k \sim \text{Poisson} \left( \frac{\lambda}{n} \right)$$

$$\text{assume that } T_k = \begin{cases} 1 & \text{if } X_k \geq 1 \\ 0 & \text{if } X_k = 0. \end{cases}$$

$$P(T_k = 1) = 1 - P(X_k = 0) = 1 - e^{-\frac{\lambda}{n}}.$$

$$E(\bar{T}_k) = 1 - e^{-\frac{\lambda}{n}},$$

$$E(\bar{T}_1 + \bar{T}_2 + \dots + \bar{T}_n) = n \left( 1 - e^{-\frac{\lambda}{n}} \right) = n \cdot (1 - e^{-\lambda/n})$$

13. Go!

$$A \text{ messages} \sim \text{Poisson}(\lambda_A)$$

$$B \text{ ---} \sim \text{Poisson}(\lambda_B)$$

words in  ${}^a$  message  $\sim P_{W|U}$ .

$$(a) \quad C \triangleq A + B \sim \text{Poisson}(\lambda_A + \lambda_B)$$

$$\begin{aligned} P(A \text{ problem}) &= P_C(q; (\lambda_A + \lambda_B) \cdot t) \\ &= e^{-(\lambda_A + \lambda_B)t} \cdot \frac{[(\lambda_A + \lambda_B) \cdot t]^q}{q!}. \end{aligned}$$

(b)  $w_i$  denotes the word length of  $i$ th message.

$$\hat{N} = W_1 + W_2 + \dots + W_C$$

$$\begin{aligned} E(N) &= E(W_1 + \dots + W_C) = E(E(W_1 + \dots + W_C | C)) \\ &= E(C \cdot E(W)) \\ &= E(C) \cdot E(W). \end{aligned}$$

$C \sim \text{Poisson}((\lambda_A + \lambda_B)t)$ ,  $W \sim \sim$  (in time duration  $t$ ).

$$E(C) = t(\lambda_A + \lambda_B), \quad E(W) = \frac{11}{6}.$$

$$E(N) = \frac{11}{6} (\lambda_A + \lambda_B) \cdot t,$$

(c) Determine the PDF of the time from  $t=0$  until receiver has received exactly eight three-word messages from A.

~~PW(3)~~ assume that  $C_3$  denotes the number of 3-word message.

each message can be calculated with prob.  $p_{W(3)}$ ,  
and  $C \sim \text{Poisson}((\lambda_A + \lambda_B)t)$ .

$C$  denotes the number of A or B message.

$$C_3 \sim \text{Poisson}((\lambda_A + \lambda_B)t \cdot p_{W(3)}).$$

~~$$P(C_3=8) = e^{-(\lambda_A + \lambda_B)t \cdot p_{W(3)}} \cdot \frac{[(\lambda_A + \lambda_B)t \cdot p_{W(3)}]^8}{8!} = P_T(t)$$~~

in duration  $t$ .

~~$$P(C_3=8) = e^{-(\lambda_A + \lambda_B)t/6} \cdot \frac{[(\lambda_A + \lambda_B)t \cdot t]^8}{8!}$$~~

$$P(C_3=8)(t) = e^{-\lambda_A t/6} \cdot \frac{(\lambda_A t \cdot t)^8}{8!}$$

(d) ① we know that each messages from ~~A~~ ~~B~~ with prob.  $\frac{\lambda_A}{\lambda_A + \lambda_B}$ , B with prob.  $\frac{\lambda_B}{\lambda_A + \lambda_B}$ .

$$P(8A \text{ and } 4B | 12 \text{ messages}) = \binom{12}{8} \left(\frac{\lambda_A}{\lambda_A + \lambda_B}\right)^8 \cdot \left(\frac{\lambda_B}{\lambda_A + \lambda_B}\right)^4.$$

$$\textcircled{2} \quad P(8A \text{ and } 4B | 12 \text{ messages}) = \frac{P(8A) \cdot P(4B)}{P(12)}$$

$$= \frac{P(A=8) \cdot P(B=4)}{P(C=12)}$$

$$= \frac{e^{-\lambda_A t} \cdot \frac{\lambda_A^8}{8!} \cdot e^{-\lambda_B t} \cdot \frac{\lambda_B^4}{4!}}{e^{-(\lambda_A + \lambda_B)t} \cdot \frac{(\lambda_A + \lambda_B)^{12}}{12!}}$$

$$= \binom{12}{8} \left(\frac{\lambda_A}{\lambda_A + \lambda_B}\right)^8 \cdot \left(\frac{\lambda_B}{\lambda_A + \lambda_B}\right)^4. \quad \square$$

14. (illuminate 透明, 照亮)

(illuminate 透明, 照亮)

$$\begin{aligned} (a) \quad E(X) &= E[X | A] \cdot P(A) + E[X | B] \cdot P(B) \\ &= E(X | A) + E(X | B) \cdot P(B) \\ &= 1 \cdot \frac{1}{2} + \frac{1}{3} \cdot \frac{1}{2} = \frac{2}{3} \end{aligned}$$

$$\begin{aligned} (b) \quad P(\text{No failures } \leq t) &= P(\min(A, B) \geq t) \\ &= P(A > t | A) \cdot P(A) + P(B > t | B) \cdot P(B) = P(A > t) \cdot P(B > t) \\ &= \frac{1}{2}(e^{-t} + e^{-3t}). \quad \begin{aligned} &\Rightarrow (e^{-t}) \cdot (e^{-3t}) \\ &= e^{-4t} \end{aligned} \end{aligned}$$

(c) N: No failures until time t.

$$P(A | N=t) = \frac{P(N=t | A) \cdot P(A)}{P(N=t)} = \frac{f_{X|A}(t) \cdot \frac{1}{2}}{f_N(t)} = \frac{\frac{1}{2}e^{-t}}{\frac{1}{2}(e^{-t} + e^{-3t})} = \frac{e^{-t}}{e^{-t} + e^{-3t}}$$

$$(d) \text{Var}(N) = \cancel{E(N^2)} - E(N)^2 \\ = E(N^2|A) \cdot P(A) + E(N^2|B) \cdot P(B) - (E(N|A) \cdot P(A) + E(N|B) \cdot P(B))$$

$$E(N^2|A) = (Var(N|A) + E(N|A)^2) \cdot P(A) + (Var(N|B) + E(N|B)^2) \cdot P(B) \\ \cancel{- E(N)}$$

$$= (1+1) \cdot \frac{1}{2} + (\frac{2}{9}) \cdot \frac{1}{2} - \frac{4}{9} \\ = \frac{2}{3}$$

$$(e) P = \left(\begin{array}{l} 1 \\ 3 \end{array}\right) \left(\frac{1}{2}\right)^{12}$$

~~(f)~~  $P = \left(\begin{array}{l} 1 \\ 4 \end{array}\right) \left(\frac{1}{2}\right)^{12}$

(g) for each bulb, <sup>life</sup> it is independent ~~between~~ among different bulbs.

$$P(N_i \leq t) = P(N_i \leq t|A) \cdot P(A) + P(N_i \leq t|B) \cdot P(B) \\ = \frac{1}{2} (1 - e^{-t} + 1 - e^{-3t})$$

$\text{Let } T_i \stackrel{\Delta}{=} N_1 + N_2 + \dots + N_i.$

~~$P(T_i \leq t) =$~~  1'

$M_{N_i}(s) = \frac{1}{2} \left( \frac{1}{1-s} + \frac{3}{3-s} \right),$

$M_{T_i}(s) = \left[ \frac{1}{2} \left( \frac{1}{1-s} + \frac{3}{3-s} \right) \right]^i, \text{ when } i=12, M_{T_{12}}(s) = \left( \frac{1}{2} \right)^{12}.$

(h)  $Q_1$ : ~~first~~ total period of first 2 B bulbs

$Q_2$ : total period of A bulb.

~~$P(Q_1 \leq t)$~~   $Q_1 \sim \text{Erlang } (\lambda=3, \text{ order}=2).$

$Q_2 \sim \text{Exponential } (\lambda=1).$

$Q_1 - Q_2 \sim ?$

$$P(Q_1 > Q_2) = ?$$

~~$$P(Q_1 > Q_2) = \frac{\lambda^k \cdot y^{k-1} \cdot e^{-\lambda y}}{(k-1)!}$$~~

~~$$f_{Q_1(x)} = q_x \cdot e^{-3x}, \quad (x \geq 0)$$~~

~~$$f_{Q_2(x)} = e^{-x}, \quad (x \geq 0)$$~~

~~$$\sum_{x=0}^{\infty} P(Q_1 > x > Q_2 | X=x) P(X=x) = P(Q_1 > Q_2)$$~~

~~$$= \sum_{x=0}^{\infty} P(Q_1 > x > Q_2) \cdot P(X=x)$$~~

$$\begin{aligned} P(Q_1 > Q_2) &= \sum_{q_2} P(Q_1 > Q_2 | Q_2 = q_2) P(Q_2 = q_2) \\ &= \int_{-\infty}^{+\infty} P(Q_1 > q) \cdot f_{Q_2}(q) dq \\ &= \int_{-\infty}^{+\infty} P(Q_2 < q) \cdot f_{Q_1}(q) dq \\ &= \int_0^{+\infty} (1 - e^{-q}) \cdot q \cdot e^{-3q} dq \\ &= \frac{7}{16}. \end{aligned}$$

(i) assume that  $\gamma_i$  denotes the period of  $i$ th bulb's illumination.  $N$  is the number of 13 bulbs in 12 bulbs  
 $T = \gamma_1 + \dots + \gamma_N$ .

$$E(T) = E(T \cdot E(N)) = 6 \cdot \frac{1}{3} = 2.$$

$$\begin{aligned} \text{Var}(T) &= E(T^2) - E(T)^2 \\ &= E(E(T^2 | N)) - E(E(T | N))^2 \\ &= E(\text{Var}(\gamma_{1N}) + E(\gamma_{1N})^2) - E(E(\gamma_{1N}))^2 \\ &= E(\text{Var}(\gamma_{1N})) + E(\text{Var}(E(\gamma_{1N}))) \\ &= E(N \cdot \text{Var}(\gamma_i)) + \text{Var}(N \cdot E(\gamma_i)) = 6 \cdot \frac{1}{9} + \frac{1}{9} \cdot \text{Var}(N) = \boxed{1} \end{aligned}$$

j)

$$P(T-t \leq x | T \geq t)$$

$$E(T-t | T \geq t) = E(T-t, A | T \geq t) + E(T-t, B | T \geq t)$$

$$= E(T-t | T \geq t, A) \cdot P(A | T \geq t) + E(T-t | T \geq t, B) \cdot P(B | T \geq t)$$

$$E(T-t | T \geq t, A) = E(T_A) = 1, \frac{1}{\lambda_A}$$

$$P(A | T \geq t) = \frac{P(T \geq t | A) \cdot P(A)}{P(T \geq t)} = \frac{P(T \geq t | A)}{P(T \geq t | A) + P(T \geq t | B)} = \frac{e^{-t}}{e^{-t} + e^{-3t}}$$

$$E(T-t | T \geq t) = \frac{e^{-t}}{e^{-t} + e^{-3t}} + \frac{1}{3} \cdot \frac{e^{-3t}}{e^{-t} + e^{-3t}}. \quad \square$$

15. Go!

16. Go! 不冲浪！相信自己！

(a)  ~~$P(N \leq n)$~~  assume that  $X_i$  means interarrival time  
 ~~$P(N=n)$~~  between two police car.

$$P(X_i \leq x) = 1 - P(X_i > x) = 1 - P(0, x) = 1 - e^{-\lambda x}.$$

$P(0, x)$  means in time  $x$ , 0 cars are noticed.

if  $X_i < \tau$ , U-turn won't happen, if  $X_i \geq \tau$ , U-turn happens.

$$P(N=n) = P(X_i < \tau)^n \cdot P(X_i \geq \tau)$$

$$= (1 - e^{-\lambda \tau})^n \cdot e^{-\lambda \tau}$$

$$E(N) = \sum_{n=1}^{+\infty} n \cdot (1 - e^{-\lambda \tau})^n \cdot e^{-\lambda \tau} = e^{\lambda \tau} - 1.$$

(b)

(b) ~~fixed~~ ~~R~~ ~~E~~  $x_1, x_2, \dots, x_n$

find (  $E(x_n | N \geq n)$  )

$$E(x_n | N \geq n) = E(x_n | x_1 < \tau, x_2 < \tau, \dots, x_{n-1} < \tau)$$

$$= E(x_n | x_n < \tau)$$

$$= \frac{\int_0^\tau x \cdot f_{x_n}(x) dx}{P(x_n < \tau)}$$

$$\underline{\underline{\int_0^\tau x}}$$

$$P(x_n < \infty) = 1 - P(0, \infty) = 1 - e^{-\lambda \infty}.$$

$$E(x_n | N \geq n) = \frac{\int_0^\tau x \cdot (\lambda + e^{-\lambda x}) dx}{1 - e^{-\lambda \tau}}$$

$$= \left( -(\tau + \frac{1}{\lambda}) \cdot e^{-\lambda \tau} + \frac{1}{\lambda} \right) / (1 - e^{-\lambda \tau}). \quad \square$$

(c)  ~~$E(x_1 + \dots + x_N) = \sum E(E(x_1 + \dots + x_N | N))$~~

~~$= \sum E(x_1 + \dots + x_N | N=n)$~~

~~$= \sum E(x_i) \cdot E(N)$~~

$$E(N) = e^{\lambda \tau} - 1, \quad E(x_i) = E(\text{Poisson}(\lambda)) = \lambda$$

$$E(x_1 + \dots + x_N + \tau) = E(E(x_1 + \dots + x_N | N)) + \tau$$

$$= \sum_{n=1}^{+\infty} [P(N=n) \cdot E(x_1 + \dots + x_N | N=n)] + \tau$$

$$= \tau + \sum_{n=1}^{+\infty} P(N=n) \cdot E(x_i | x_i \leq \tau) \cdot n$$

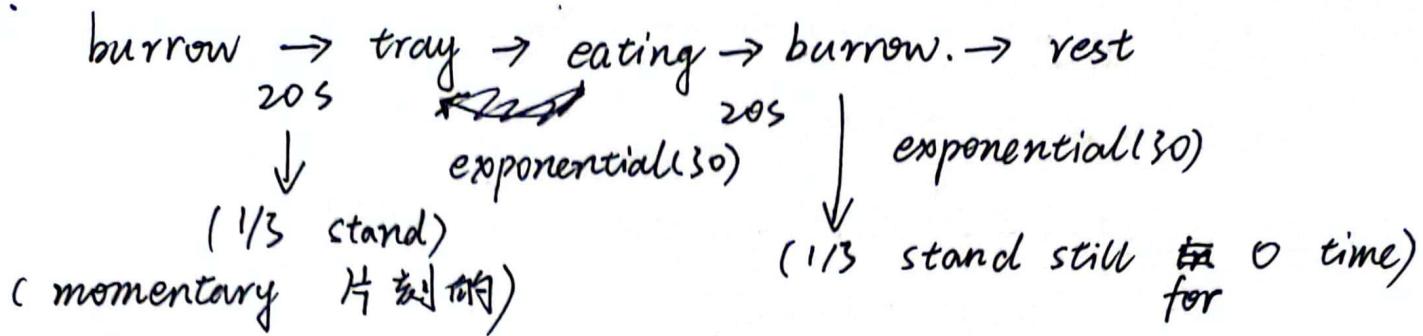
$$= \tau + E(N) \cdot E(x_i | N \geq i)$$

$$= \tau + (e^{\lambda \tau} - 1) \cdot \left( \frac{\frac{1}{\lambda} - (\tau + \frac{1}{\lambda}) \cdot e^{-\lambda \tau}}{1 - e^{-\lambda \tau}} \right)$$

$$= \frac{1}{\lambda} (e^{\lambda \tau} - 1). \quad \square$$

完成

17.



assume that  $x_i$  denotes the ~~i-th~~ time of finishing  $i$ th cycle.

~~$$x_i = t_{i,1} + t_{i,2} + t_{i,3} + t_{i,4}$$~~

$$t_{i,1} = t_{i,3} = 20,$$

$$t_{i,2}, t_{i,4} \sim \text{Exponential}(30).$$

assume that photographer arrive at  $c$ .

$$P(c \text{ in } t_{i,1}) = P(c \text{ in } t_{i,3}) = \frac{2}{5},$$

We can discard  $t_{i,3}$  and  $t_{i,4}$  without loss of generality.

$$x_i = t_{i,1} + t_{i,2}.$$

assume that photographer arrive at 0 time unit.

Waiting time is  $T$ ,  $A$  means snapping the bear.

~~$$\begin{aligned} E(T) &= E(T|A) \cdot P(A) + E(T|A^c) \cdot P(A^c) \\ &= 10 \cdot \frac{1}{3} + (50 + E(Y)) \cdot \frac{2}{3}, \end{aligned}$$~~

~~$$E(T) = 110.$$~~

$Z$  is the time he waits totally.

~~assume that~~  $c$  is the time photographer arrives.

~~$$E(Z) = E(Z|c \text{ in } t_{i,1}) P(t_{i,1}) + E(Z|c \text{ in } t_{i,2}) P(t_{i,2})$$~~

~~$$E(Z|c \text{ in } t_{i,2}) = 30 + E(T) = 140$$~~

~~$$E(Z|c \text{ in } t_{i,1}) =$$~~

~~$$E(Z|T_{i,1}, B) \cdot P(B|T_{i,1}) + E(Z|T_{i,1}, B^c) \cdot P(B^c|T_{i,1})$$~~

$B$  denotes bear stand still.

$$E(Z|T_{i,1}) = 10 \cdot \frac{1}{3} + \frac{2}{3} \cdot (20 + 30 + E(T)) = 110.$$

$$\begin{aligned} E(Z) &= \frac{2}{5} \cdot 110 + \frac{3}{5} \cdot 140 \\ &= 128. \end{aligned}$$

$$E(T) = E(T|A) \cdot P(A) + E(T|A^c) \cdot P(A^c)$$

$$= 10 \cdot \frac{1}{3} + (50 + E(Y)) \cdot \frac{2}{3}$$

$$\Rightarrow E(T) = 110.$$

$$E(Z|T_{i,2}) = 30 + E(T) = 140$$

$$E(Z) = E(Z|T_{i,1}) \cdot P(T_{i,1}) + E(Z|T_{i,2}) \cdot P(T_{i,2})$$

$$E(Z|T_{i,1}) = \int_0^{20} E(Z|T_{i,1}, c) \cdot f_{Z|T_{i,1}}(c) dc$$

$$= \int_0^{20} \left( E(Z|T_{i,1}, c, B) \cdot P(B|T_{i,1}, c) + E(Z|T_{i,1}, c, B^c) \cdot P(B^c|T_{i,1}, c) \right) dc$$

$$= \int_0^{20} \left( \frac{(20-c)}{2} \cdot \frac{1}{3} \cdot \frac{20-c}{20} + (20-c)(1 - \frac{1}{3} \cdot \frac{20-c}{20}) \cdot \frac{1}{20} \right) dc$$

$$= 140 - \frac{40}{3} - \frac{10}{9}$$

$$E(Z) = (140 - \frac{40}{3} - \frac{10}{9}) \cdot \frac{2}{5} + 140 \cdot \frac{3}{5}$$

$$= 140 - \frac{16}{3} - \frac{4}{9} \approx 134.22. \quad \square$$

18. Go! 呀，看来我不先开口，哪个虫儿敢伴声。

$P(k, t)$  means in  $t$  time interval  $[0, t]$ , exactly exists  $k$  arrivals.

$$P(\underbrace{\text{arrivals}}_{\leq k} | \underbrace{\text{arrivals}}_{\leq m}) = \frac{P(\text{arrivals} \leq k) - P(\text{arrivals} \leq m)}{P(\text{arrivals} \leq m)}$$

$$\begin{aligned}
 P(T \leq x | \text{1 arrival in } [0, t]) &= \frac{P(\cancel{\text{1}}, x) \cdot P(0, t-x)}{P(1, t)} \\
 &= \frac{e^{-\lambda x} \cdot (\lambda x) \cdot e^{-\lambda(t-x)}}{e^{-t\lambda} \cdot (t\lambda)} \\
 &= \frac{x}{t}.
 \end{aligned}$$

$T | \text{1 arrival in } [0, t] \sim \text{Uniform}(0, t)$ .  $\square$

19.  $X_1 \sim \text{Exponential}(\cancel{\lambda}), X_2 \sim \text{Exponential}(\lambda_2)$ .

$$(a) P(\max\{X_1, X_2\} \leq x) = P(X_1 \leq x) \cdot P(X_2 \leq x)$$

$$\begin{aligned}
 E(\max\{X_1, X_2\}) &= \int_0^{+\infty} x \cdot [(1 - e^{-\lambda_1 x}) \cdot \lambda_1 e^{-\lambda_2 x} + (1 - e^{-\lambda_2 x}) \cdot \lambda_2 e^{-\lambda_1 x}], \\
 \Rightarrow f_{\max\{X_1, X_2\}}(x) &= \frac{dP(\max\{X_1, X_2\} \leq x)}{dx} = f_{X_1}(x) \cdot f_{X_2}(x) + F_{X_1}(x) \cdot f_{X_2}(x)
 \end{aligned}$$

$$E(\max\{X_1, X_2\}) = E(X_1) + E(X_2) - E(\text{exponential}(\lambda_1 + \lambda_2))$$

$$= \frac{1}{\lambda_1} + \frac{1}{\lambda_2} - \frac{1}{\lambda_1 + \lambda_2}$$

$$= \frac{\lambda_1^2 + \lambda_2^2 + \lambda_1 \cdot \lambda_2}{\lambda_1 \cdot \lambda_2 (\lambda_1 + \lambda_2)}.$$

$$\begin{aligned}
 (b) f_{T, Z}(x) &= f_T(x) \cdot F_Z(x) + F_T(x) \cdot f_Z(x) \\
 &= (\lambda_1 \cdot e^{-\lambda_1 x}) \cdot (1 - e^{-\lambda_2 x} - e^{-\lambda_2 x} \cdot \lambda_2 x) + (1 - e^{-\lambda_1 x}) \cdot \lambda_2^2 x \cdot \\
 &= \lambda_1 \cdot e^{-\lambda_1 x} + \lambda_2 \cdot e^{-\lambda_2 x} + e^{-(\lambda_1 + \lambda_2)x} \cdot (-\lambda_1 - \lambda_1 \lambda_2 x - \lambda_2^2 x)
 \end{aligned}$$

$$E(\max\{T, Z\}) = \cancel{\frac{1}{\lambda_1} + \frac{1}{\lambda_2}} - \frac{\lambda_1}{(\lambda_1 + \lambda_2)^2} - \frac{3\lambda_2}{(\lambda_1 + \lambda_2)^2}. \quad \square$$

20.  $T_k \sim \text{Erlang}(k)$ ,

$$f_{T_k}(y) = \lambda^k y^{(k-1)} \cdot e^{-\lambda y} / (k-1)!$$

$$\begin{aligned} \sum_{k=1}^{\infty} f_{T_k}(y) &= \sum_{k=1}^{\infty} \lambda^k y^{(k-1)} \cdot e^{-\lambda y} / (k-1)! \\ &= \sum_{k=1}^{\infty} \frac{\lambda \cdot (\lambda y)^{(k-1)} \cdot e^{-\lambda y}}{(k-1)!} \\ &= \lambda \cdot e^{-\lambda y} \cdot \sum_{k=0}^{\infty} \frac{(\lambda y)^k}{k!} \\ &= \lambda \cdot e^{-\lambda y} \cdot e^{\lambda y} = \lambda. \quad \square \end{aligned}$$

21.  $x_1(k)$ : time of  $k$ th arrival in Poisson( $\lambda_1$ ) Process.  
 $x_2(k)$ : time of  $k$ th arrival in Poisson( $\lambda_2$ ) Process.

Consider  $x_1 + x_2$  obeys Poisson( $\lambda_1 + \lambda_2$ ) distribution,

For each arrival, it is from  $x_1$  with  $\frac{\lambda_1}{\lambda_1 + \lambda_2}$  probability.

$P(x_1(n) < x_2(m))$  means when  $x_1$  has  $n$  arrivals,  $x_2$  must less than or equal to  $m-1$  arrivals.

It's the same as in first  $(n+m-1)$  trials,  $x_1 \geq n$ .

if  $x_1 + x_2 = n+m-1$ ,  $x_1 \geq n \iff x_1(n) > x_2(m)$ .

$$\begin{aligned} P(x_1(n) < x_2(m)) &= P(x_1 \geq n \mid x_1 + x_2 = n+m-1) \\ &= \sum_{k=n}^{n+m-1} \binom{n+m-1}{k} \cdot \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^k \cdot \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{n+m-1-k}. \quad \square \end{aligned}$$

$$22. Y = X_1 + X_2 + \dots + X_N$$

$N \geq 0$ ,  $X_i \sim \text{Bernoulli}(p)$

(a)  $Y = X_1 + X_2 + \dots + X_N$ ,  $N \sim \text{Binomial}(n, p)$ .

$N$  can be explained that in  $m$  trials,  $\overset{N}{\underset{\text{X}}{\exists}}$  = the number of success with probability  $p$ .

and  $Y$  = the number of success with probability  $p$  in  $m$  trials.

$Y \sim \text{Binomial}(m, p)$ .

(b) if  $N \sim \text{Poisson}(\lambda)$

$N$  can be seen as an interval  $[0, 1]$ , splitted in to  $n$  intervals equally. ( $n \rightarrow \infty$ ). each interval is a Bernoulli distribution with prob.  $\lambda/n$ .  $M$  = the number of all ~~chosen~~ points chosen being the

and  $T$  denotes that on each ~~not~~ chosen points, it has prob.  $p$  having value 1, that is ~~not~~ chosen with prob.  $p \cdot \lambda/n$ .

$Y \sim \text{Poisson}(\lambda p)$ .  $\square$

$$23. X_i \sim \text{Exponential}(\lambda), Y = X_1 + \dots + X_N$$

$N \sim \text{Geometric}(p)$ .

$X_i$  is the time of the  $i$ th arrival in  $\text{Poisson}(\lambda)$ .

$X_1 + \dots + X_N$  denotes the first accepted arrival with probability  $p$ , that is, the first accepted arrival with probability  $\lambda p$ .

$Y \sim \text{Exponential}(\lambda p)$ .  $\square$

## 24. $T \sim \text{Exponential}(\nu)$

$N = \underline{\text{number of Poisson process in } [0, T]}$ .

$x_i \triangleq \text{the } i^{\text{th}} \text{ Poisson process. } (\lambda)$   
 $\text{time of}$

$$P(N=n) = P(x_1 + \dots + x_n \leq T, x_{n+1} + \dots + x_n > T).$$

that is, when  $T$  first success,  $N$  is exactly ~~not~~ having  $n$  arrivals. assume that  $H$  is the number of  $\text{Poisson}(\lambda) + \text{Poisson}(\nu)$

$$\cancel{P(N=n)} = P(H \cancel{= n}, H=n | H=n+1) \sim \text{Erlang}(n+1, \nu+\lambda) \\ \cancel{Z(n+1)}.$$

$$\therefore P(N=n) = \cancel{P(\text{exactly } n+1, \nu+\lambda)} \cdot \cancel{P\left(\frac{\nu}{\nu+\lambda}\right) \cdot \left(\frac{\lambda}{\nu+\lambda}\right)^n}$$

$$\cancel{P(n+1, \nu+\lambda)} Z(n+1) = 1 - \sum_{i=0}^n P(i, \nu+\lambda) \\ = 1 - \left( \sum_{i=0}^n e^{-(\nu+\lambda)} \cdot \frac{(\nu+\lambda)^i}{i!} \right)$$

$$P(N=n | Z(n+1)) = \left(\frac{\nu}{\nu+\lambda}\right) \cdot \left(\frac{\lambda}{\nu+\lambda}\right)^n. \quad ? \quad ? \quad ?$$

## 25. Ans!!! $\checkmark$ An infinite <sup>server</sup> ~~severe~~ queue.

customer arrives  $\sim \text{Poisson}(\lambda)$  rate.

$x_i = \text{customer stay}$

$$x_i \in \{1, 2, \dots, n\}$$

?  $N_t = \text{number of customers at time } t$ . (assume that  $t \geq n$ )

$$E(N_t) = \cancel{E(E(N_t | N))} = \sum_{n=1}^{\infty} E(N_t | N) \cdot P(N=n)$$

$$E(N_t | N=n) = \cancel{P(\cancel{n, \lambda})} E(\text{Poisson}(n\lambda)) = n\lambda.$$

$$E(N_t) = E(N \cdot \lambda) = \lambda \cdot E(N) = \lambda \cdot E(x_i). \quad \cancel{N \triangleq x_i}. \quad \square$$

$$26. P(X_1=x_1, X_2=x_2) = P(X_1+x_2=x_1+x_2, X_1=x_1)$$

$$\downarrow X_1 \sim \text{Poisson}(\lambda p), X_2 \sim \text{Poisson}(\lambda(1-p))$$

$$P(X_1=x_1, X_2=x_2) = e^{-\lambda} \cdot \frac{\lambda^{x_1+x_2}}{(x_1+x_2)!} \cdot \binom{x_1+x_2}{x_1} \cdot p^{x_1} \cdot (1-p)^{x_2}$$

$$= e^{-\lambda} \cdot p^{x_1} \cdot (1-p)^{x_2} \cdot \frac{\lambda^{x_1} \cdot \lambda^{x_2}}{x_1! \cdot x_2!} \cdot p^{x_1} \cdot (1-p)^{x_2}$$

$$= e^{-\lambda p} \cdot p^{x_1} \cdot \lambda^{x_1} \cdot \frac{1}{x_1!} \cdot e^{-\lambda(1-p)} \cdot \frac{(\lambda(1-p))^{x_2}}{x_2!}$$

$$= P(X_1=x_1) \cdot P(X_2=x_2). \square$$

$$27. \cancel{F(E=e)} = F(L=e+E)$$

$$\cancel{F(L_2=l)} = \cancel{F(L_1=l - E(\text{Poisson}(\frac{2}{\lambda})))}$$

$$= \tilde{F}(L_2 - L_1 + L_1 = l)$$

$h_1(L_2 - L_1)$  is the time from  $t$  to the first time ~~of arrival~~  
since the memorylessness of Poisson process,  $h_1(L_2 - L_1)$  obeys  
Poisson( $\frac{2}{\lambda}$ ) distribution,  $L_1$  is a Poisson( $\frac{2}{\lambda}$ ) distribution.

Symmetrically,  $(L_2 - L_1) - h_1(L_2 - L_1)$  is also Poisson( $\frac{2}{\lambda}$ ) distribution.

$$L_2 \sim \text{Erlang}(3, \frac{\lambda}{2}). \quad \square$$