

Poisson Process

1) Renewal Process:

Def: i) $\varphi = \{t_n\}_{n \geq 1}$ is simple point process. st.

$$0 < t_1 < t_2 < \dots < t_n < \dots, \quad t_n \rightarrow \infty \text{ as } n \rightarrow \infty.$$

ii) $\{N(t)\}_{t \geq 0}$ is counting process. st.

$$N(t) = \max \{n \mid t_n \leq t\}.$$

Rmk: $X_n = t_n - t_{n-1}$ is n^{th} interval time.

Def: A point process $\varphi = \{t_n\}_{n \geq 1}$ is renewal process if $(X_n)_n = (t_n - t_{n-1})_n$ is i.i.d.

Thm. For renewal process $\varphi = \{t_n\}_{n \geq 1}$. We have:

$$\begin{cases} \lim_{t \rightarrow \infty} N(t)/t = \lambda, \text{ w.p. 1.} \\ \lim_{t \rightarrow \infty} E[N(t)]/t = \lambda. \end{cases} \quad \text{where } \lambda = 1/E(X_1).$$

Pf: 1) By SLLN: $\sum_{k=1}^{N(t)} X_k / N(t) \leq t / N(t) \leq \sum_{k=1}^{N(t)+1} X_k / N(t).$

2) prove: $N(t)/t$ is u.i.

Truncate: $\hat{X}_n = a I_{\{X_n \geq a\}} \leq X_n. \therefore \hat{N}(t) \geq N(t)$

where a is chosen: $p(X_n \geq a) \triangleq p \in (0, 1)$

Note: Arrival occur only at $na, n \in \mathbb{Z}^+$.

Denote k_n is number of arrivals at na .

(i.e. contain the arrival spent time $< a$)

$$\therefore k_n \sim \text{Geop}.$$

(Times spent for success = arrival times)

$$\therefore N(t) \leq \tilde{N}(t) \leq \sum_{i=1}^{\lceil t/\tau \rceil} k_i = S(t). \quad E(S(t)) = O(t).$$

$$\begin{aligned} \therefore E(N(t)/t^2 I_{\{N(t) \geq t\}}) &\leq E(S(t)/t^2) P(\frac{N(t)}{t} \geq 1) \\ &\leq \frac{A}{c^2} \rightarrow 0. \quad (\text{Chebyshev}) \end{aligned}$$

(2) Poisson Point Process:

Def: Poisson process with rate $\lambda \in (0, \infty)$ is a renewal process $\varphi = \{t_n\}_{n \in \mathbb{Z}^+}$ s.t. $X_n \sim \text{Exp}(\lambda)$.

① Exponential Dist.:

i) It's memoryless: $P(X > a+b | X > a) = P(X > b)$
for $a, b \geq 0$.

Thm. Nonnegative, nondegenerated r.v. X is memoryless $\Leftrightarrow X \sim \text{Exp}(\lambda)$.

Pf: Set $g(x) = P(X > x)$. $\therefore g(x+y) = g(x)g(y)$

$$\therefore g(n) = g(1)^n, \quad g(\frac{n}{m}) = g(\frac{1}{m})^n = g(1)^{\frac{n}{m}}$$

$$\text{Approx. 'R' by 'Q': } g(x) = g(1)^x$$

Link: For discrete case:

$$P(X > n+k | X > k) = P(X > n), \quad \forall n, k \in \mathbb{Z}^+.$$

$$\Leftrightarrow X \sim \text{Geop}(p), \quad \text{for } p \in (0, 1).$$

$$\text{Pf: } P(X > n) = P(X > 1)^n \stackrel{a}{=} p^n$$

$$\Rightarrow P(X = n) = p^{n-1}(1-p)$$

ii) Approx:

Thm. $X_n \sim \text{Geo}(p_n)$. $\lim_n np_n = \lambda \Rightarrow X_n/n \xrightarrow{d} X \sim \text{Exp}(\lambda)$.

$$\begin{aligned} \text{pf: } \varphi_{X_n/n}(t) &= \frac{p_n}{1 - (1-p_n)e^{it/n}} \\ &= \frac{np_n}{(1 - e^{it/n})/1/n + np_n e^{it/n}} \longrightarrow \frac{\lambda}{\lambda - it} \end{aligned}$$

iii) n-folds:

Thm. $t_n = \sum_{k=1}^n X_k \sim \text{Gamma}(n, \lambda)$. if $X_k \sim \text{Exp}(\lambda)$. i.i.d.

$$\text{pf: } f = \lambda e^{-\lambda t}, \quad f^{*n} = \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!}$$

iv) Combination:

prop. For $X_1 \sim \text{Exp}(\lambda_1)$, $X_2 \sim \text{Exp}(\lambda_2)$. $Z = \min(X_1, X_2)$.

$$(a) \quad Z \sim \text{Exp}(\lambda_1 + \lambda_2)$$

$$(b) \quad Z | X_1 < X_2 \sim Z | X_1 > X_2 \sim Z.$$

$$(c) \quad p(X_1 > X_2) = \lambda_1 / (\lambda_1 + \lambda_2), \quad p(X_1 < X_2) = \lambda_2 / (\lambda_1 + \lambda_2)$$

rkmk: Z means X_1, X_2 work simultaneously.

⑤ Poisson Approx:

Thm. $\{X_{n,k}\}_{1 \leq k \leq n}$ indep. $p(X_{n,k}=1) = p_{n,k}$, $p(X_{n,k}=0) = 1 - p_{n,k}$

$$S_n = \sum_{k=1}^n X_{n,k}. \quad \text{If } \sum_{k=1}^n p_{n,k} \rightarrow \lambda \in (0, \infty) \text{ as } n \rightarrow \infty, \text{ and}$$

$$\max_{1 \leq k \leq n} p_{n,k} \rightarrow 0 \quad \text{Then } S_n \xrightarrow{d} \text{Poisson}(\lambda).$$

Pf: $\varphi_{S_n}(t) = \prod_1^n (1 + p_{n,k} (e^{it} - 1))$

$$\ln (1 + p_{n,k} (e^{it} - 1)) = p_{n,k} (e^{it} - 1) + \theta_k |p_{n,k} (e^{it} - 1)|^2$$

where $|\theta_k| \leq 1$. Since for large n : $|p_{n,k} (e^{it} - 1)| \leq \frac{1}{2}$

check $\sum_1^n \ln (1 + p_{n,k} (e^{it} - 1)) \rightarrow \lambda (e^{it} - 1)$.

Cor. For $\{X_{n,k}\}$ indep. nonnegative. $p(X_{n,k}=1) = p_{n,k}$.

$$p(X_{n,k} \geq 2) = \sum_k p_{n,k}^2. \quad S_n = \sum_k^n X_{n,k}.$$

If $\sum p_{n,k} \rightarrow \lambda$. $\max_k p_{n,k} \rightarrow 0$. $\sum_k^n p_{n,k}^2 \rightarrow 0$. Then

$$S_n \rightarrow_d \text{Poisson}(\lambda).$$

Pf: Truncate: $X'_{n,k} = X_{n,k} \mathbb{I}_{\{X_{n,k} \leq 1\}} \leq X_{n,k}$

$$\therefore S'_n = \sum X'_{n,k} \rightarrow_d \text{Poisson}(\lambda).$$

check $\forall \varepsilon > 0$. $p(|S_n - S'_n| \geq \varepsilon) \rightarrow 0$ ($n \rightarrow \infty$)

(3) Characterization:

Def: i) Point process $\varphi = \{t_n\}$ has stationary increment if $N(t+s) - N(t)$ indep with t , but on s .

ii) Point process $\varphi = \{t_n\}$ has indep increment if

$$\forall I_1, I_2 = \emptyset. \quad N(I_1) \text{ indep with } N(I_2).$$

① Thm. For $\varphi = \{t_n\}$ point process with $N(t)$:

i) It's indep increment

ii) It's stationary increment.

iii) It's sparse: $p(N(0,h)=1) = \lambda h + o(h)$. $p(N(0,h) \geq 2) = o(h)$

Then $N(0, t) \sim \text{Poisson}(\lambda t)$

Pf. Denote $X_t = N(0, t) = \sum_{k=1}^r (X_{\frac{r}{n}t} - X_{\frac{(k-1)}{n}t}) \stackrel{\Delta}{=} \sum_k X_{nk}$

Check μ_{nk}, Σ_{nk} satisfies approxi. condition.

Then let $n \rightarrow \infty, \therefore X_t \sim \text{Poisson}(\lambda t)$.

Rmk: Directly, $P(N(t) \leq k) = P(t_k \geq t)$

$$= \int_t^{\infty} \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!} dt = e^{-\lambda t} (1 + \lambda t + \dots)$$

if $\{t_n\}$ is poisson process with rate λ .

Thm. (little o.c.t.)

If $\{t_n\}$ is poisson process. Then $P(N(t) > 0) = \lambda t + o(t)$.

$P(N(t) > 1) = o(t), \lambda = E(N(1)), t \rightarrow 0$

Pf. $P(N(t) > 0) = 1 - e^{-\lambda t} = \lambda t + o(t)$

$P(N(t) > 1) = e^{-\lambda t} (\frac{(\lambda t)^2}{2} + \dots) = o(t)$.

Rmk: Check: $N(t+s) - N(t) \sim N(s)$. And by indep of $\{X_n = t_n - t_{n-1}\}$. So a Poisson process also satisfies characterization i), ii), iii).

(Partition by arrival time $\{t_n\}$, $I \cap \{t_n\} = \emptyset$?)

② Generating Func.

Define: $G_{N(t)}(z) = E(z^{N(t)})$, $|z| \leq 1$. g.f of

$N(t)$ counting process.

$$\text{Note: } \frac{1}{\Delta t} (G_{N(t+\Delta t)} - G_{N(t)}) =$$

$$\frac{1}{\Delta t} E(z^{N(t)} (z^{N(t+\Delta t) - N(t)} - 1))$$

$$\stackrel{\text{indep}}{=} \frac{1}{\Delta t} E(z^{N(t+\Delta t) - N(t)} - 1) E(z^{N(t)})$$

$$= \frac{1}{\Delta t} E(z^{N(\Delta t)} - 1) E(z^{N(t)})$$

$$E(z^{N(\Delta t)} - 1) = p(N(\Delta t)=0) - 1 + p(N(\Delta t)=1)z + \sum_{k \geq 2} p(N(\Delta t)=k)z^k$$

$$1') \quad p(N(\Delta t)=0) = p(N(\Delta t)=0) p(N(\Delta t-\Delta s)=0)$$

$$\Rightarrow \exists \lambda. \quad p(N(\Delta t)=0) = e^{-\lambda \Delta t}$$

$$2') \quad p(N(\Delta t)=1)z + \sum_{k \geq 2} p(N(\Delta t)=k)z^k$$

$$= p(N(\Delta t)=1) \left(z + \sum_{k \geq 2} \frac{p_k}{p_1} z^k \right)$$

$$\left| \sum_{k \geq 2} \frac{p_k}{p_1} z^k \right| \leq \sum_{k \geq 2} \frac{p_k}{p_1} |z|^k \leq \frac{p(N(\Delta t) \geq 2)}{p(N(\Delta t)=1)}$$

$\rightarrow 0 \quad (\Delta t \rightarrow 0)$ by sparsity (little occ.)

$$\Rightarrow LHS = \frac{1}{\Delta t} (e^{-\lambda \Delta t} - 1) + G_{N(t)} \frac{(z + o(\Delta t)) p_1}{\Delta t}$$

$$\rightarrow (\lambda z - \lambda) G_{N(t)}(z). \quad (\Delta t \rightarrow 0)$$

$$\text{i.e. we obtain: } G'_{N(t)}(z) = (\lambda z - \lambda) G_{N(t)}(z)$$

$$\Rightarrow G_{N(t)}(z) = G_{N(0)}(z) e^{-(\lambda + \lambda z)t} = e^{-(\lambda + \lambda z)t}$$

Rmk: Commonly, let $\lim_{\Delta t \rightarrow 0} p(N(\Delta t)=1)/\Delta t = \lambda$.

Modification:

i) Ignore "Stationary increment":

$$\text{Then: } E(z^{N(t+\Delta t) - N(t)} - 1) \neq E(z^{N(\Delta t)} - 1)$$

$$\bar{E} e^{z \frac{\mu(t+\Delta t) - \mu(t)}{\Delta t} - 1} = p(\Delta N=0) - 1 + \sum_{k \geq 1} p(\Delta N=k) z^k.$$

ALL assumption: $\lim_{\Delta t \rightarrow 0} \frac{p(\mu(t+\Delta t) - \mu(t)=0) - 1}{\Delta t} = -\lambda(t).$

$$\Rightarrow \lim_{\Delta t \rightarrow 0} p(\Delta N=1)/\Delta t = \lambda(t). \quad \int_0^t \lambda(s) ds \Rightarrow G_{\mu(t)}(z) = e^{(z-1) \int_0^t \lambda(s) ds}.$$

ii) Ignore "Sparsity":

Assume: $p(N(\Delta t)=k)/p(N(\Delta t)=1) \xrightarrow{\Delta t \rightarrow 0} p_k$. $p(N(\Delta t) \geq 1)/\Delta t \xrightarrow{\Delta t \rightarrow 0} \lambda$.

$$\text{Then: } \frac{1}{\Delta t} \bar{E} e^{z \frac{\mu(t+\Delta t) - \mu(t)}{\Delta t} - 1} = \frac{p(N(\Delta t)=0) - 1}{\Delta t} + \frac{p(N(\Delta t) \geq 1)}{\Delta t} \left(\sum_{k \geq 1} \frac{p(N(\Delta t)=k)}{p(N(\Delta t) \geq 1)} z^k \right)$$

$$\xrightarrow{\Delta t \rightarrow 0} -\lambda + \lambda \sum_{k \geq 1} p_k z^k$$

$$\Rightarrow G_{\mu(t)}(z) = e^{\lambda t (\sum_{k \geq 1} p_k z^k - 1)}$$

iii) Ignore "Indept increment":

Consider compound Poisson process $Y_t = \sum_{k=1}^{N(t)} X_k$.

1) If $X_k \stackrel{i.i.d.}{\sim} F_X$ indept with $\mu(t)$:

$$G_{Y(t)}(z) = \bar{E} \left(\bar{E} e^{z \sum_{k=1}^{N(t)} X_k} \mid N(t)=n \right) = \bar{E} (G_{X(t)}(z))$$

$$= e^{\lambda (G_X(z) - 1)t}$$

2) To remove "indept". Consider $Y_t = \sum_{k=1}^{N(t)} X_k(t, S_k)$

S_k is r.v. the k^{th} arrival time, which is called filtering Poisson process. Calculate its ch.f:

$$\phi_{Y(t)}(z) = \bar{E} e^{iz Y(t)},$$

$$= \bar{E} \bar{E} e^{iz \sum_{k=1}^n X_k(t, S_k)} \mid S_1, \dots, S_n, N(t)=n$$

Denote: $B_k(t, S_k) = E_{X_k} e^{iz X_k(t, S_k)},$

$$\Rightarrow LHS = E_N(E_S(\prod_{i=1}^n B_k(t, s_k) \mid N(t)=n))$$

$$= E_N(\int_{0 \leq s_1 \leq \dots \leq s_n \leq t} \frac{n!}{t^n} \prod_{i=1}^n B_k(t, s_k) \lambda s_1 \dots \lambda s_n)$$

$$\stackrel{\text{csgm}}{=} E_N(\frac{1}{n!} \int_{[0,t]^n} \frac{n!}{t^n} \prod_{i=1}^n B_k(t, s_k) \lambda \vec{s})$$

$$= E_N(\lambda \int_0^t B_k(t, s) \lambda s ds) = e$$

follows from $(s_k)_i \mid N(t)=n \stackrel{N(t)}{\sim} \{X_k\}_i$ order stat.

(4) Partition:

① Bernoulli Trails:

Consider at time t_n . Do a "Bernoulli Trail", which generates object of type 1 with prob. p type 2 with prob $1-p$ indeptly in $\mathcal{U} = \{t_n\}$ Poisson Process.

Thm. If $X \sim \text{Poisson}(\alpha)$, X_1 is number of objects of type 1. X_2 is number of type 2. Then $X_1 + X_2 = X$. $X_1 \sim \text{Poisson}(p\alpha)$. $X_2 \sim \text{Poisson}((1-p)\alpha)$. X_1, X_2 indept.

p.f. To show: X_1, X_2 are Poisson. dist.

$$\text{Note: } p(X_1=k, X_2=m) = p(X_1=k, X=k+m)$$

$$= p(X_1=k \mid X=k+m) p(X=k+m)$$

$$= \frac{(p\alpha)^k}{k!} e^{-p\alpha} \frac{(q\alpha)^m}{m!} e^{-q\alpha}$$

where $q \stackrel{\Delta}{=} 1-p$.

Thm. $\psi = \{ \psi_i \}$. λ -Poisson process. ψ_i is point process of type i arrivals. $i=1, 2$. Then ψ_i is λ_i -Poisson process. ψ_1 is λ_1 -Poisson process. ψ_2 is λ_2 -Poisson process. indpt each other.

Pf: ψ_i are poisson process by charac. and the parameter is from $X = X_1 + X_2$. Thm above.

For indpt. consider $(A_i)_1^n, (B_j)_1^m$ two collections of disjoint intervals.

Show: $(X_1(A_1) \dots X_1(A_n))$ indpt of $(X_2(B_1) \dots X_2(B_m))$

For A_i : $X_1(A_i \cap B_j)$ indpt of X_2 by partition indptly

$X_1(A_i - B_j)$ indpt of X_2 by indpt increments

$\Rightarrow X_1(A_i) = X_1(A_i \cap B_j) + X_1(A_i - B_j)$ indpt with $X_2(B_j)$

for $\forall j \in \{1, 2, \dots, m\}$.

(2) Superposition:

We can put $\psi_1 \sim \text{Poi}(\lambda_1), \psi_2 \sim \text{Poi}(\lambda_2)$ indpt processes

together to obtain $\psi = \psi_1 + \psi_2$. So $N_t = N_{1,t} + N_{2,t}$.

$\Rightarrow \psi \sim \text{Poisson}(\lambda_1 + \lambda_2)$. by charac.

Besides, denote $Y_i \sim \text{exp}(\lambda_i)$, the arrival time of type

$i, i=1, 2$. Then the partition prob of type 1 is:

$P(Y_1 < Y_2) = \lambda_1 / (\lambda_1 + \lambda_2)$. of type 2 is $\lambda_2 / (\lambda_1 + \lambda_2)$.

(5) Construct a Poisson Process:

Fix $t > 0$. Consider $u_k \stackrel{i.i.d.}{\sim} U[0, t]$, $1 \leq k \leq n$.

1) Note: γ is Poisson (λ) Process. for set:

$$p(t_1 \leq s \mid N(t) = 1) = \frac{p(N(s) = 1, N(t) - N(s) = 0)}{p(N(t) = 1)} = \frac{s}{t}$$

2) Recall: $(u_1, \dots, u_n) \sim \mathcal{U} = n!/t^n$, order stat.

$$\text{prop. } (t_1, t_2, \dots, t_n) \mid N(t) = n \stackrel{\mathcal{U}}{\sim} (u_1, \dots, u_n)$$

for $\mathcal{U} = [t, \infty)$ poisson process.

$$\text{pf: } \text{Note: } \{t_i = s_i, 1 \leq i \leq n, N(t) = n\} =$$

$$\{X_1 = s_1, X_2 = s_2 - s_1, \dots, X_n = s_n - s_{n-1}, X_{n+1} > t - s_n\}$$

where $X_i \stackrel{i.i.d.}{\sim} \text{exp}(\lambda)$.

\Rightarrow To simulate a Poisson process up to time t :

First simulate the value of $N(t)$. if $N(t)$

$= n$. then: generate n i.i.d $U[0, t]$, r.v.'s:

(u_1, \dots, u_n) , place them in (u_1, \dots, u_n) .

(6) Application:

Def: $M/G/k$ is a queue model:

i) "M" means Markovian, i.e. modulated by Poisson Process.

ii) "G" means service time has general distribution.

iii) "k" means there're k servers.

Next, consider $M/G/\infty$: Arrivals time $\{t_n\} \sim \text{Poi}(\lambda)$.

Service times $S_n \stackrel{i.i.d.}{\sim} \text{Exp}(\mu)$ with mean μ^{-1}

Define: $X(t)$ is number of customers in system at time t .

prop. $X(t) \sim \text{Poisson}(\alpha(t))$, $\alpha(t) = \int_0^t \lambda p(s > x) dx$.

Rmk: Since $\alpha(t) \xrightarrow{t \rightarrow \infty} \int_0^\infty \lambda p(s > x) dx = \lambda/\mu =: \rho$.

$$\Rightarrow \lim_{t \rightarrow \infty} P(X(t) = n) = e^{-\rho} \rho^n / n!. \quad \forall n \geq 0.$$

Pf: n^{th} customer arrives at t_n . Departs at $t_n + s_n$.

Define $D(t)$ is number of departs at time t .

$\Rightarrow N(t) = X(t) + D(t)$ partitioned in 2 types.

Suppose u is r.v. of arrival time. S is service time.

A customer still in system at time $t \Leftrightarrow$

$$u + s > t. \Leftrightarrow s > t - u.$$

$S_0 = p(t) = p(s > t - u)$ is partition prob.

$u \sim u(0, t)$ since we know he arrives in $[0, t]$.

$$\Rightarrow t - u \sim u(0, t). \therefore p(t) = p(s > u) = \int_0^t \frac{1}{t} p(s > x) dx.$$

$$S_0 : \alpha(t) = \lambda t p(t) = \int_0^t \lambda p(s > x) dx.$$

Rmk: Partition in binomial trials: $(\text{under } N(t) = n)$

$$X(t) = \sum_{i=1}^n I_{\{s_i > t - u_i\}} = \sum_{i=1}^n I_{\{s_i > t - u_i\}}.$$

Consider "stay" is "1". "Depart" is "0".

Then $X(t) | N(t) = n \sim \text{Bin}(n, p(u + s > t))$