

Applied Probability 2015

1) ca) Birth Death Chain: Let $S = \mathbb{Z}_{\geq 0}$ be state space,

$Q: S \times S \rightarrow \mathbb{R}$ defined by: ~~$Q_{m,n}$~~

$$Q(n, m) = \begin{cases} \lambda_n & m = n+1 \\ \mu_n & m = n-1 \\ (-\lambda_n - \mu_n) & m = n \\ 0 & |n-m| > 1 \end{cases}$$

$$(\lambda_n, \mu_n \geq 0)$$

Let $(X_t)_{t \geq 0}$ be continuous markov chain with Q as a generator:

$(X_{nt})_{t \geq 0}$ is a birth death chain

ca) If π is a solution to Detailed Balance Equations (DB):

$$\forall n \in S: \sum_{m \geq 0} \pi_m Q_{m,n} = \sum_{m \geq 0} \pi_n Q_{n,m} = \pi_n \sum_{m \geq 0} Q_{n,m} = 0$$

$\therefore \pi$ is invariant.

If π is invariant: $(n \neq m) \quad Q_{n,m} \neq 0$ iff $|n-m|=1$

$$\therefore \text{(DBE): } \forall n \in S, \quad \pi_n \cdot Q_{n,n+1} = \pi_{n+1} \cdot Q_{n+1,n} \quad (*)$$

$$n=0: \sum_{m \geq 0} \pi_m Q_{m,0} = -\pi_0 \cdot \lambda_0 + \pi_1 \mu_1 \stackrel{(*)}{=} \text{True for } n=1$$

If $(*)$ True for some $N-1$:

$$\sum_{m \geq 0} \pi_m Q_{m,N} = \pi_N (-\lambda_N - \mu_N) + \pi_{N-1} \lambda_{N-1} + \pi_{N+1} \mu_{N+1} = 0$$

$$\text{cf } (*): \pi_N \overset{\mu}{\lambda}_N = \pi_{N-1} \lambda_{N-1} \Rightarrow \pi_N \lambda_N = \pi_{N+1} \overset{\mu}{\lambda}_{N+1}$$

\therefore Induction complete $\Rightarrow \pi$ satisfies DB.

cb) $S = \mathbb{Z}_{\geq 0} : (X_t)_{t \geq 0}$ is a Birth-Death Chain,

$$\lambda_n = \lambda, \quad \mu_n = \min\{n, s\} \cdot \mu$$

Chain is clearly irreducible. Let $(Y_n)_{n \geq 0}$ be jump chain.

$\therefore (X_t)_{t \geq 0}$ is recurrent iff $(Y_t)_{t \geq 0}$ is recurrent iff
 state s is recurrent (in $(Y_n)_{n \geq 0}$) iff $IP(\text{return to } s \mid Y_0 = s) = 1$

But $IP \neq \alpha$

Claim: $\alpha = IP(\text{hit } s \mid Y_0 = s-1) = 1$; Consider jump chain with
 modification s.t. $S \rightarrow S-1$ with $IP=1$.
 \therefore This does not affect α .

$\{1, \dots, s\}$ is a finite, irreducible markov chain \Rightarrow
 $\alpha = 1$

Claim: $\beta = IP(\text{hit } s \mid Y_0 = s+1) = 1$ iff $\lambda \leq S \cdot \mu$. Consider
 modification s.t. $S \rightarrow S+1$ with $IP=1$ (will not change β)

$\therefore \{s, \dots, k : k \geq s\}$ is state irreducible class; $(Y_n)_{n \geq 0}$ on this
 class is a random walk; $\therefore IP(+1) = \lambda / \lambda + S\mu$, $IP(-1) = S\mu / \lambda + S\mu$
 \therefore Recurrent iff $\lambda \leq S\mu \Rightarrow IP(\text{hit } s \mid \beta = 1)$ iff $\lambda \leq S\mu$

$\therefore IP(\text{return to } s \mid Y_0 = s) = 1$ iff $\lambda \leq S\mu$.

If $\lambda \leq S\mu$: Chain is recurrent & Non-explosive.
 \therefore +ve invariant iff invariant distribution exists.

6 ca): Suffice to solve invariant D.B.

$$\begin{aligned} \therefore \pi_n \cdot \lambda &= \pi_{n+1} \cdot \mu \cdot (n+1) & (n+1 \leq s) \\ &= \pi_{n+1} \cdot \mu \cdot s & (n+1 > s) \end{aligned}$$

$$\begin{aligned} \therefore \pi_k &= \frac{(\lambda/\mu)^k}{k!} \pi_0 & (0 \leq k \leq s) \\ \pi_{s+k} &= \pi_s \cdot (\lambda/\mu_s)^k \end{aligned} \quad \left. \begin{array}{l} \sum \pi_k < \infty \text{ iff} \\ \sum (\lambda/\mu_s)^k < \infty \text{ iff } \lambda < \mu \cdot s \end{array} \right\}$$

\therefore $\lambda > S \cdot \mu$ (transient)
 $\lambda = S \cdot \mu$ (null-recurrent)
 $\lambda < S \cdot \mu$ (+ve-recurrent)

If π is X -invariant: $\pi \cdot P = \pi$

$$Q = \lambda \cdot P - \lambda I \quad (I \text{ is identity matrix})$$

$$\therefore \pi \cdot Q = \lambda [\pi \cdot P - \pi] = 0$$

$\therefore \pi$ is $(Y_t)_{t \geq 0}$ invariant.

(c) $T_a = \sum_{k=1}^{T_a} X_k$, $(X_k)_{k \geq 1}$ is iid $\text{Exp}(1)$ R.V. (generating $(N_t)_{t \geq 0}$)

$(X_t)_{t \geq 0}$: $X_t = k$ for $t \in [X_1 + \dots + X_k, X_1 + \dots + X_{k+1})$
 $(X_k)_{k \geq 1}$ independent of X

$$\therefore E[T_a] = E[E[T_a | T_a = n]] = E[E[\sum_{k=1}^n X_k | T_a = n]]$$

$$(X_n)_{n \geq 1} \text{ independent of } T: E[T_a] = E[E[n \cdot 1]] = E[X_1] \cdot E[n \sim T_a]$$

$$\therefore E[T_a] = E[T_a] \cdot \lambda$$

2.) (a) Let $(X_t)_{t \geq 0}$ be a r.n.g. process with independent increments;

$$\left. \begin{aligned} IP(X_{t+h} - X_t = 0) &= 1 - \lambda h + o(h) \\ IP(X_{t+h} - X_t = 1) &= \lambda h + o(h) \end{aligned} \right\} \text{Uniformly for all } t \in \mathbb{R}_+$$

$(X_t)_{t \geq 0}$ is a Poisson process of rate λ .

Let $X_t = N_t + M_t$: N_t, M_t have independent increments \Rightarrow
 X_t has independent increments.

$$\begin{aligned} IP(X_{t+h} - X_t = 0) &= IP(N_{t+h} - N_t = M_{t+h} - M_t = 0) = (1 - \lambda h + o(h)) \cdot (1 - \mu h + o(h)) \\ &= 1 - (\lambda + \mu)h + o(h) \end{aligned}$$

$$\begin{aligned} IP(X_{t+h} - X_t = 1) &= IP\left(\left\{N_{t+h} - N_t = 1, M_{t+h} - M_t = 0\right\}\right) + \\ &\quad IP\left(\left\{N_{t+h} - N_t = 0, M_{t+h} - M_t = 1\right\}\right) \\ &= (\lambda h + o(h))(1 - \mu h + o(h)) + (\mu h + o(h))(1 - \lambda h + o(h)) \\ &= (\lambda + \mu)h + o(h) \end{aligned}$$

$\therefore (X_t)_{t \geq 0}$ is a Poisson process of rate $\lambda + \mu$.

(b) At state x : IP transition to $y = P(x, y)$

Let $q_x = r_x = P(x, x)$: We reject only move out of state x at a rate of λ , thinned by probability p .

\therefore Exit rate $= (1-p) \cdot \lambda$

$$IP(X \rightarrow y) = \frac{P(x, y)}{1-p} \quad (y \neq x)$$

Q matrix: Let S be state space;

$$\left. \begin{aligned} (x \neq y) \quad Q(x, y) &= P(x, y) \cdot \lambda \\ Q(x, x) &= -\lambda(1 - P(x, x)) \end{aligned} \right\} Q \text{ is generator of } (X_t)_{t \geq 0}$$

3.) (a) Infinitesimal Definition: Let $(X_t)_{t \geq 0}$ be a sequence of $\text{Poi}(cp)$ E.V. independent of X_t . ^{iid}

Since $(X_t)_{t \geq 0}$ has independent increments, $(Y_t)_{t \geq 0}$ has independent increments as well.

$$\begin{aligned} \mathbb{P}(Y_{t+h} - Y_t = 0) &= \mathbb{P}(X_{t+h} - X_t = 0) + \mathbb{P}(X_{t+h} - X_t = 1, \text{reject}) \\ &= (1 - \lambda h + o(h)) + (1-p)(\lambda h) + o(h) \\ &= 1 - (\lambda p)h + o(h) \end{aligned}$$

$$\begin{aligned} \mathbb{P}(Y_{t+h} - Y_t \geq 1) &= \mathbb{P}(Y_{t+h} - Y_t \geq 1, \text{accepts}) + \mathbb{P}(Y_{t+h} - Y_t \geq 1, \text{reject}) \\ &\leq \mathbb{P}(X_{t+h} - X_t \geq 2) = o(h) \end{aligned}$$

$$\therefore \mathbb{P}(Y_{t+h} - Y_t = 1) = (\lambda p)h + o(h)$$

$\therefore \text{Poi}(\lambda \cdot p \text{ rate})$ process

$$\mathbb{P}(X_t - Y_t = n, Y_t = m) = \mathbb{P}(X_t = n+m, m \text{ accepts out of } n+m)$$

$$= \frac{e^{-\lambda t} \lambda^{n+m}}{(n+m)!} \frac{(n+m)!}{n! m!} p^n (1-p)^m = \frac{e^{-\lambda p t} (\lambda p)^m}{m!} \frac{e^{-\lambda(1-p)t} (\lambda(1-p))^n}{n!}$$

$$= \mathbb{P}(X_t - Y_t = n) \cdot \mathbb{P}(Y_t = m)$$

$\therefore (X_t - Y_t, Y_t)$ independent (fixed t)

(b) By applying (a) repeatedly to n bins: $(M_n)_{n \geq 0}$ # of balls in Bin k
 $(Y_t^k)_{t \geq 0}$ is $\text{Poi}(\lambda/n)$, $(Y^k : 1 \leq k \leq n)$ are independent.

$$\mathbb{P}(M_n \geq d) = \mathbb{P}(\text{Poi}(\lambda/n) \geq d)^n$$

At time n : $Y_n^k \sim \text{Poi}(1)$

$$\therefore \mathbb{P}(M_n \geq d) = \mathbb{P}(\text{Poi}(1) \geq d)^n \leq e^{-n \{d - d \log d\}}$$

$$\text{If let } \log(n) = x : n \left\{ d - d \log d \right\} = e^x \left\{ (1+\epsilon) \frac{x}{\log x} (1 - \log(1+\epsilon) \frac{\log x}{\log x}) \right\}$$

$$= e^x \frac{x(1+\epsilon)}{\log x} \left\{ 1 - \log(1+\epsilon) - \log(x) + \log(\log(x)) \right\} \quad (*)$$

$$\text{But } e^x \cdot x(1+\epsilon) \rightarrow +\infty \text{ as } x \rightarrow \infty;$$

$$\frac{1 - \log(1+\epsilon) - \log(x) + \log(\log(x))}{\log(x)} =$$

$$-1 + \frac{1 - \log(1+\epsilon)}{\log(x)} + \frac{\log(\log(x))}{\log(x)}$$

$\rightarrow 0 \text{ as } x \rightarrow \infty$
 $\quad \quad \quad \leftarrow x \rightarrow \infty \Rightarrow \log(x) \rightarrow \infty$

$$= -1 + \lim_{x \rightarrow \infty} \frac{\log(\log(x))}{\log(x)} = -1 + \lim_{x \rightarrow \infty} \frac{1}{x} = -1$$

$$\therefore (*) \rightarrow -\infty$$

$$\therefore 0 \leq \lim_{n \rightarrow \infty} \mathbb{P} \left(M_n \geq (1+\epsilon) \frac{\log n}{\log(\log n)} \right) = 0.$$

$$\leq \lim_{x \rightarrow \infty} e^x \frac{x(1+\epsilon)}{\log x} \left\{ 1 - \log(1+\epsilon) - \log(x) + \log(\log(x)) \right\}$$

$$= 0$$

$$\therefore \lim_{n \rightarrow \infty} \mathbb{P} \left(M_n \geq (1+\epsilon) \frac{\log n}{\log(\log n)} \right) = 0.$$

$\hat{\varepsilon}$ density: $f(x) = x \cdot \lambda^2 e^{-\lambda x}$

$$E[\hat{\varepsilon}] = \lambda^2 \int_0^{\infty} x^2 e^{-\lambda x} dx = \lambda^2 \left(\int_0^{\infty} x e^{-\lambda x} dx + \int_0^{\infty} x e^{-\lambda x} dx \right) \\ = 2 E[\varepsilon] = 2/\lambda$$

$$\lim_{t \rightarrow \infty} E[A_t - A_1 | \varepsilon] = 2/\lambda$$

$\{e_1, \dots, e_N\}$

cc) Q - matrix: $Q(x, x + e_i - e_j) = \lambda_j P_{ji}$ ($x_j \geq 1$)

$$Q(x, x) = - \sum_{k=1}^N 1_{x_k > 0} \lambda_k x_k$$

R - matrix ($N=1$): $Q(e_i, e_j) = \lambda_j P_{ji}$

~~$\lambda_j Q(e_i, e_j) = \lambda_j P_{ji} Q(e_j, e_i)$~~

~~Invariant Equations: π_j~~

~~$\pi_j \lambda_j P_{ji} = \sum_{i=1}^N \pi_i \lambda_i P_{ij} P_{ji}$~~

Since there are finit states, invariant measure π exists

$P_{ji} > 0 \Rightarrow \forall e_i, e_j: \lambda_j Q(e_i, e_j) > 0 \Rightarrow$ Irreducible

Irreducible, finit state jump chain \Rightarrow Recurrent $\Rightarrow (X_t)_{t \geq 0}$

recurrent \Rightarrow invariant measure is unique (up to constant)

Finit states $\Rightarrow \sum \pi < \infty \Rightarrow \exists$ Unique invariant distribution

cd) $\pi(N) = \prod_{i=1}^N \frac{\lambda_i^{n_i}}{n_i!}$

~~$\sum_N \pi(N) Q(N, N) =$~~

$$\sum_{x \in \mathbb{Z}^d} \frac{1}{\min\{1, \|x\|_\infty^d\}} < \infty \quad (1)$$

If $d \leq 0$: $\max_{x \in \mathbb{Z}^d} \{1, \|x\|_\infty^d\} = 1 \Rightarrow (1) = \infty$

Let $d > 0$: (1) iff $\sum_{x \in \mathbb{Z}^d \setminus \{0\}} \frac{1}{\|x\|_\infty^d} = \sum_{n \geq 1} \{n^2 - (n-1)^2\} \cdot \frac{1}{n^d} \quad (d \geq 1)$

$$= \sum_{n \geq 1} \frac{2n-1}{n^d} < \infty \quad \text{iff} \quad d > 2$$

(1) iff $\sum_{n \geq 1} \frac{2}{n^d} < \infty \quad \text{iff} \quad d > 1$

\therefore the recurrent iff $d > d_c$

2.) (a) $M(\lambda) / M(\mu) / \infty$ Queue: Let $S = \mathbb{Z}_{\geq 0}$ be state space, $(X_t)_{t \geq 0}$ is a $(M(\lambda) / M(\mu) / \infty)$ queue if it is generated by Q matrix:

$$Q(n, m) = \lambda \cdot \mathbb{1}_{m=n+1} + (\mu n) \mathbb{1}_{m=n-1} + \mathbb{1}_{n=m} (-\lambda - \mu n)$$

Invariant: $\pi_n = e^{-(\lambda/\mu)} \cdot (\lambda/\mu)^n / n!$

Burke's process: Let $(D_t)_{t \geq 0}$ be the departure process.
 $(D_t)_{t \geq 0}$ is a Poisson rate λ process.

$\forall T \geq 0$: X_T is independent of $(D_s)_{0 \leq s \leq T}$

(b) Since departure process is a Poisson process: We can model it as a renewal process, inter-renewal time of E_n , $(E_n)_{n \geq 1}$ is iid $\text{Exp}(\lambda)$ R.V.

$A_2^{(t)} - A_1(t)$ distribution \rightarrow Size biased distribution of E

Applied Probability 2013

1.) (a) Consider the jump chain: $(Y_n)_{n \geq 0}$

$$P(n, n+1) = \frac{1}{1 + \varepsilon_n}, \quad P(n, n) = \frac{\varepsilon_n}{1 + \varepsilon_n}$$

Let $h_k = IP((Y_n)_{n \geq 0} \text{ returns to } 0 \mid Y_0 = k)$: $h_0 = 1$,
 $h_k \in [0, 1]$

Recurrence: $h_k = \frac{\varepsilon_k}{1 + \varepsilon_k} + \frac{1}{1 + \varepsilon_k} h_{k+1}$

$$(1 + \varepsilon_k) \cdot h_k = \varepsilon_k + h_{k+1} \Rightarrow (h_k - h_{k+1}) = \varepsilon_k (1 - h_k) \geq 0$$

$\therefore (h_k)_{k \geq 0}$ is decreasing sequence.

Since Markov chain is irreducible: Chain is transient iff state 0 is transient iff state 0 is jump-chain transient

$$IP(\exists n \geq 1 \text{ s.t. } Y_n = 0 \mid Y_0 = 0) = 1 - \prod_{k=0}^{\infty} \frac{1}{1 + \varepsilon_k} \quad (*)$$

If $\sum_{k=0}^{\infty} \varepsilon_k < \infty$: $\varepsilon_k \rightarrow 0 \Rightarrow \exists N \in \mathbb{N}$ s.t. $k \geq N \Rightarrow \log(1 + \varepsilon_k) \geq \varepsilon_k/2$

$$\therefore \prod_{k=0}^{\infty} \frac{1}{1 + \varepsilon_k} \geq \prod_{k=0}^{N-1} \frac{1}{1 + \varepsilon_k} \cdot e^{-\sum_{k=N}^{\infty} \varepsilon_k/2} > 0$$

$\therefore (*) < 1 \Rightarrow 0$ is transient $\Rightarrow (X_t)_{t \geq 0}$ is transient.

If $\sum_{k=0}^{\infty} \varepsilon_k = \infty$: Let $h_k = IP((Y_n)_{n \geq 0} \text{ returns to } 0 \mid Y_0 = k)$, $h_0 = 1$
 $h_k \in [0, 1]$

$$h_k - h_{k+1} = \varepsilon_k (1 - h_k) \geq 0 \Rightarrow (h_k)_{k \geq 0} \text{ decreasing.}$$

$$\therefore h_k - h_{k+1} \geq \varepsilon_k \cdot (1 - h_1) \quad (h_1 \geq h_k)$$

$$\therefore h_1 - h_{k+1} \geq (1 - h_1) \sum_{r=1}^k \varepsilon_r \rightarrow \infty \text{ if } h_1 \neq 1 \text{ (reject!)} \quad \text{if } h_1 = 1$$

$\therefore h_{k+1} \geq h_1 = 1 \Rightarrow$ Chain is recurrent at 0
 \Rightarrow Recurrent.

(b) Chain is transient: (A.S.) $\exists \mathbb{N}$ s.t. $\forall Y_N = 0, Y_{N+k} = k$

$$\therefore \sum_{k=1}^N \frac{1}{\lambda_{k-1}(1+E_{k-1})} + \sum_{k=1}^{\infty} \frac{Z_{N+k}}{(\epsilon_{k-1}+1)\lambda_{k-1}} \quad (Z_k)_{k \geq 1} \text{ i.i.d. Exp}(1)$$

\therefore Let $(J_n)_{n \geq 0}$ be jump kernel

Birth chain

$$J = J_N + \sum_{k \geq 1} \frac{Z_k}{\lambda_{k-1}(1+E_{k-1})}, \quad (Z_k)_{k \geq 1} \text{ i.i.d. Exp}(1) \text{ r.v.}$$

Consider a birth only chain, $q(n, n+1) = \lambda_n (1+E_n)$

A.S. Explosive iff $\sum_{n \geq 1} \frac{1}{\lambda_n (1+E_n)} < \infty$

$\zeta < \infty$ iff $\sum_{k \geq 1} \frac{Z_k}{\lambda_{k-1}(1+E_{k-1})} < \infty$ iff ζ is the total time of birth chain as described above.

\therefore Criteria: $\sum_{k \geq 0} \frac{1}{\lambda_k (1+E_k)} < \infty$ iff A.S. Explosive.

(c) $\pi_n = \pi_{n-1} \cdot \lambda_{n-1} \quad (n \neq 0) \Rightarrow \pi_0 = 1, \pi_n = \prod_{k=0}^{n-1} \lambda_k$

$$\pi_0 = \sum_{n \geq 1} \pi_n (\lambda_n E_n)$$

Invariant equation: $\pi_n \cdot \lambda_n (1+E_n) = \pi_{n-1} \cdot \lambda_{n-1} \quad (n \geq 1)$

$$\therefore \pi_n = \pi_{n-1} \frac{\lambda_{n-1}}{\lambda_n (1+E_n)} = \pi_0 \prod_{k=0}^{n-1} \frac{\lambda_k}{\lambda_{k+1} (1+E_k)} = \frac{\lambda_0}{\lambda_n} \prod_{k=0}^{n-1} \frac{1}{1+E_k} \pi_0$$

$$(\pi_0 \lambda_0) \pi_0 = \sum_{k \geq 1} \pi_k \cdot \lambda_k E_k = \pi_0 \cdot \lambda_0 \underbrace{\sum_{k \geq 1} \frac{E_k}{(1+E_0) \dots (1+E_{k-1})}}_{=1}$$

$\therefore \pi_n$ invariant!

Special case: $\pi_n = \pi_0 \prod_{k=0}^{n-1} \frac{1}{1+\frac{d}{n+1}} = \pi_0 \prod_{k=0}^{n-1} \frac{n+1}{n+1+d}$

If $d \leq 1$: $\prod_{k=0}^{n-1} \frac{n+1}{n+1+d} \geq \prod_{k=0}^{n-1} \frac{n+1}{n+2} = \frac{1}{n+1} \Rightarrow \sum \pi_n \geq \sum \frac{1}{n+1} = \infty$

\therefore No invariant distribution \Rightarrow Not +ve recurrent

If $d > 1$: $\sum \varepsilon_n = \infty \Rightarrow$ Recurrent \Rightarrow Non-explosive

General stirling approx.: $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$

$$(n+d)! \sim \sqrt{2\pi(n+d)} \left(\frac{n+d}{e}\right)^{n+d}$$

$$\therefore \sim \sum \frac{1}{n^a} < \infty$$

\therefore Invariant ^{Distribution} Measure Exists, (Suspicious)

2) ca) Poisson Process: Let $(\xi_n)_{n \geq 1}$ be iid $\text{Exp}(\lambda)$ R.V.,
 $J_0 = 0, J_n = \sum_{k=1}^n \xi_k$

Define $X_t = k$ for $t \in [J_k, J_{k+1})$: $(X_t)_{t \geq 0}$ is a Poisson process.

Q-matrix Definition: $X_0 = 0$; $Q(n, n+1) = \lambda, Q(n, n) = -\lambda,$
 $Q(n, m) = 0 \quad (m \neq n, n+1)$

Infinitesimal: $(X_t)_{t \geq 0}$ has independent increments and

$$\left. \begin{aligned} \mathbb{P}(X_{t+h} - X_t = 0) &= 1 - \lambda h + o(h) \\ \mathbb{P}(X_{t+h} - X_t = 1) &= \lambda h + o(h) \end{aligned} \right\} \text{Uniformly in } t \text{ as } h \rightarrow 0$$

Poisson: $(X_t)_{t \geq 0}$ has stationary, independent increments.
 $\forall t \geq 0, X_t \sim \text{Poi}(\lambda t)$

cb) $P_k(t) = \mathbb{P}(X_t = k)$

$$P_k(t+h) = \sum_{i=0}^k P_i(t) \mathbb{P}(X_{t+h} - X_t = k-i)$$

$$\begin{aligned} &= P_k(t) (1 - \lambda h + o(h)) + P_{k-1}(t) (\lambda h + o(h)) \quad (k \geq 1) \\ &= P_k(t) (1 - \lambda h + o(h)) \quad (k=0) \end{aligned}$$

$$\frac{P_k(t+h) - P_k(t)}{h} = -\lambda(t) P_k(t) + \sum_{k \neq 0} \lambda(t) P_{k-1}(t) + o(h)$$

~~manipulate~~

$\therefore P_k(t)$ is cont; Right derivative exist

$$t \rightarrow t-h: \text{Left derivative exist, } = \lim_{h \rightarrow 0} -\lambda(t-h) P_k(t-h) + \sum_{k \neq 0} \lambda(t-h) P_{k-1}(t-h)$$

$$= -\lambda(t) \{P_k(t) - P_{k-1}(t)\}$$

$$\therefore P_k'(t) = \lambda(t) \{-P_k(t) + P_{k-1}(t)\} \quad [k \neq 0]$$

$$P_0'(t) = -\lambda(t) P_0(t) \Rightarrow P_0(t) = A e^{-\lambda(t)}, \quad P_0(0) = 1 \Rightarrow A = 1$$

$$\text{Let } G(t, z) = \sum_{k \geq 0} P_k(t) z^k.$$

$$\frac{\partial G}{\partial t} = \sum_{k \geq 0} \{-\lambda(t) P_k(t) + \sum_{k \neq 0} P_{k-1}(t) \lambda(t)\} z^k$$

$$= -\lambda(t) G(t, z) + \lambda(t) z G(t, z)$$

$$\therefore G_t = \lambda(t) G \{z - 1\} \Rightarrow G(t, z) = A e^{(z-1) \int_0^t \lambda(s) ds}$$

$$\text{Given } G(0, z) = 1 \Rightarrow A = 1$$

$$\therefore \text{MGF of } X(t) = \text{MGF of } \text{Poi}(\Lambda(t)) \quad \text{R.V.}$$

$$\therefore X(t) \sim \text{Poi}(\Lambda(t))$$

(c) Fix t : Let $(Y_n)_{n \geq 1}$ be subsequence of X , $(Y_n)_{n \geq 1}$ are elements $> t$; $(Z_n)_{n \geq 1}$ be elements $\leq t$

$\therefore (R_s)_{0 \leq s \leq t}$ is a function of $(Z_n)_{n \geq 1}$, $(R_s)_{t \leq s < \infty}$ is a function of $(Y_n)_{n \geq 1}$

$\therefore (R_s)_{0 \leq s \leq t}, (R_s)_{t \leq s < \infty}$ are independent.

$\therefore R_{t_1} - R_{t_2}$ independent of $(R_s)_{0 \leq s \leq t_2}$ ($t_1 > t_2 \geq 0$)
 \therefore Independent increments.

$$\begin{aligned} \mathbb{P}(X_{t+h} - X_t = 0) &= \mathbb{P}(Y_1 \geq t+h) = \frac{\int_{t+h}^{\infty} f(x) dx}{\int_t^{\infty} f(x) dx} \\ &= \frac{1 - F(t+h)}{1 - F(t)} \end{aligned}$$

$$= 1 + \frac{F(t) - F(t+h)}{1 - F(t)} = 1 - \underbrace{h \frac{f(t)}{1 - F(t)}}_{\lambda(t)} + o(h) \quad (\text{Mean-value theorem})$$

$$\begin{aligned} \mathbb{P}(X_{t+h} - X_t \geq 2) &\leq \mathbb{P}(Y_1, Y_2 \in [t, t+h]) = \left\{ \frac{\int_t^{t+h} f(x) dx}{\int_t^{\infty} f(x) dx} \right\}^2 \\ &= \left(\frac{h f(t)}{1 - F(t)} + o(h) \right)^2 = o(h^2) \end{aligned}$$

$$\begin{aligned} \therefore \mathbb{P}(X_{t+h} - X_t = 1) &= 1 - \mathbb{P}(X_{t+h} - X_t = 0) + o(h) \\ &= h \cdot \lambda(t) + o(h) \end{aligned}$$

$$\therefore (b): R_t \sim \text{Poi}(\Lambda(t))$$

$$\Lambda(t) = \int_0^t \frac{f(s)}{1 - F(s)} ds = - \left[\log(1 - F(s)) \right]_0^t = -\log(1 - F(t))$$

$$\therefore \text{Poi}(-\log(1 - F(t)))$$

Proof of hint:

$$Y_n: \begin{cases} p & (t) \\ 1-p & (t) \end{cases} \left. \begin{array}{l} f(t)/1-F(t) \\ f(t)/F(t) \end{array} \right\}$$

We redefine the process $(X_n)_{n \geq 1}$:

Let $(\xi_n)_{n \geq 1}$ be iid $\text{Ber}(0, p)$ R.V., $p = F(t)$

$(d_n)_{n \geq 1}$ be iid, density $f(t)/F(t)$

$(b_n)_{n \geq 1}$ be iid, density $f(t)/(1-F(t))$

$$\begin{aligned} X_n &= a_n & \text{if } \xi_n = 1 \\ &= b_n & \text{if } \xi_n = 0 \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \begin{array}{l} \text{Same distribution as original} \\ (X_n) \end{array}$$

$$\hookrightarrow \mathbb{P}(Y_1 \in A_1, \dots, Y_n \in A_n) = \mathbb{P}(d_1 \in A_1, \dots, d_n \in A_n).$$

$$\sum_{n_1, \dots, n_n} \mathbb{P}(Y_1 = X_{n_1}, \dots, Y_n = X_{n_n}) \cdot \underbrace{\mathbb{P}(d_{n_1} \in A_1, \dots, d_{n_n} \in A_n)}_{= \mathbb{P}(d_1 \in A_1, \dots, d_n \in A_n)}$$

$$= \mathbb{P}(d_1 \in A_1, \dots, d_n \in A_n)$$

(Similarly for Z_n)

$$\mathbb{P}((Y_1, \dots, Y_n) \in A, (Z_1, \dots, Z_n) \in B) = \sum_{i,j} \mathbb{P}((Y_1, \dots, Y_n) = (X_{i_1}, \dots, X_{i_n}), (Z_1, \dots, Z_n) = (X_{j_1}, \dots, X_{j_n})) \cdot \mathbb{P}((d_{i_1}, \dots, d_{i_n}) \in A, \mathbb{P}(d_{j_1}, \dots, d_{j_n}) \in B)$$

$$= \mathbb{P}((d_1, \dots, d_n) \in A) \cdot \mathbb{P}((b_1, \dots, b_n) \in B)$$

4) (a) M/M/1: ~~Arrivals follow a $Poi(\lambda)$ process~~ and each
 Let $S = \mathbb{Z}_{\geq 0}$: Q-matrix: $Q(n, n+1) = \lambda$, $n \geq 0$
 $Q(n, n-1) = \mu$, $n \geq 1$
 $Q(n, n) = -(\lambda + \mu)$

$(X_t)_{t \geq 0}$ be the process generated by Q : $(X_t)_{t \geq 0}$ is a
 (M, M, 1) Queue.

Consider Jump chain: $(\frac{\lambda}{\lambda+\mu}, \frac{\mu}{\lambda+\mu})$ walk on S
 \therefore Recurrent iff $\mu/\lambda + \mu \geq 1/2$ iff $\mu \geq \lambda$

$\therefore (X_t)_{t \geq 0}$ is recurrent iff $\lambda \leq \mu$ (Recurrent \Rightarrow ^{Non-}Explosive)

Invariant Distribution: $\pi_n \cdot \lambda = \pi_{n+1} \cdot \mu$ (Detailed Balance)

$\therefore \pi_n = (1 - \lambda/\mu) (\lambda/\mu)^n$ is an invariant measure;
 (Distribution) iff $(\lambda/\mu) < 1$

\therefore +ve recurrent iff $\lambda < \mu$.

(λ, μ)

Burke's Theorem: If $(X_t)_{t \geq 0}$ is a M/M/1 Queue, +ve recurrent and
 in equilibrium: Let $(D_t)_{t \geq 0}$ be departure process
 $D_t \sim \text{Poi}_{\lambda}^{\text{SSN}}$ process of rate λ ; $X_{t\epsilon}$ is
 independent of $(D_s)_{0 \leq s \leq t}$ ($\forall t \geq 0$)

(Proof): Let $(\hat{X}_t)_{t \geq 0}$, $(\hat{X}_s)_{s \in [0, t]}$ be time reversal

Since we solved for detailed balance, $(X_s)_{0 \leq s \leq t}$ is

time reversible $\Rightarrow (\hat{X}_s)_{s \in [0, t]}$, $(X_s)_{0 \leq s \leq t}$

have the same distribution

But arrivals of \hat{X} corresponds to departures of X .

\therefore Departure process of $(X_s)_{0 \leq s \leq t}$ has the same
 distribution as arrival process of $\hat{X} \Rightarrow \text{Poisson } \lambda$ process

$\therefore X_0$ is independent of arrival process of $(X_s)_{0 \leq s \leq t}$

$\therefore X_t$ independent of departure process $(D_s)_{0 \leq s \leq t}$

(b) Jackson's Theorem: Let $\tilde{\lambda}_0 = \sum_{j \neq 0} \tilde{\lambda}_j P_{ji} + \lambda_i$ be a ≥ 0 system of linear equations. \exists unique solutions to $\tilde{\lambda}_0$

If $\tilde{\lambda}_0 / \mu_0 < 1$: ($1 \leq i \leq n$)

Queue system is in equilibrium with invariant is the recurrent with invariant distribution $\pi_{pr} = \prod_{k=1}^n (1 - \tilde{\lambda}_0 / \mu_k) (\tilde{\lambda}_0 / \mu_k)^{n_k}$

In equilibrium: Departure process (from queue to outside world) are independent poisson processes of rate $\tilde{\lambda}_i \cdot P_{i,0}$.

$$(P_{i,0} = \sum_{j \neq i} -P_{ij} + 1)$$

Process (L_1, \dots, L_n) is not independent:

$IP(\text{no } L_1, \dots, L_{n-1} = 0 \text{ on } [0,1], \text{ no arrival to } L_n \text{ on } [0,1], \text{ extncel})$

$1 \text{ arrival to } L_n \text{ on } [0,1]) = 0$

But $IP(L_1, \dots, L_{n-1} = 0 \text{ on } [0,1]) > 0$, $IP(\text{no extncel arrival, 1 intncel arrival to } L_n \text{ on } [0,1]) > 0$

\therefore Not independent

Let $(\hat{L}_1, \dots, \hat{L}_n)_{0 \leq s \leq t}$ be reversed process at equilibrium

$$\hat{Q}_{n, n+e_i-e_j} = \frac{\pi_{n+e_i-e_j}}{\pi_n} Q_{n+e_j-e_i} \quad (n_j \geq 1)$$

$$= \frac{(\tilde{\lambda}_i/\mu_i)}{(\tilde{\lambda}_j/\mu_j)} P_{i,j} \mu_i$$

$$= P_{i,j} (\tilde{\lambda}_i/\tilde{\lambda}_j \mu_i)$$

$$\hat{Q}_{n, n+e_i} = (\tilde{\lambda}_i/\mu_i) P_{i,0} \mu_i = P_{i,0} \tilde{\lambda}_i$$

$$\hat{Q}_{n, n-e_i} = 1/(\tilde{\lambda}_i/\mu_i) \lambda_i = \lambda_i \mu_i / \tilde{\lambda}_i$$

$\therefore (\hat{L}_1, \dots, \hat{L}_n)$ is also a Jackson Network

Departure process of $(L_1, \dots, L_n)_{0 \leq s \leq t} \equiv$ Arrival process of

$(\hat{L}_1, \dots, \hat{L}_n) \Rightarrow$ Independent Poisson process

L_i departure process =

$\mathbb{P} \{ \hat{L}_i \text{ arrival} \} \equiv \text{Poisson } P_{i,0} \tilde{\lambda}_i$

$L_t = \hat{L}_0$, \hat{L}_0 independent of arrivals in $(L_{0,s})_{0 \leq s \leq t}$.

$\therefore \mathbb{P} \{ L_t \text{ independent of departures in } 0 \leq s \leq t. \}$

No.:

Date:

Applied Probability 2014

1.) (a) Let $(\varepsilon_n)_{n \geq 1}$ be iid $\text{Exp}(1)$ R.V. ; S be state space,
 $\forall s \in S, q(s) = -Q_{ss}$

Define $P: S \times S \rightarrow [0,1], P(a,b) = Q(a,b)/q(a);$

Let $(Y_n)_{n \geq 0}$ be discrete markov chain with transition matrix $P, Y_0 = X_0.$

Define : $T_0 = 0, T_n = \sum_{k=1}^n \varepsilon_k / q(Y_{k-1}) :$

Let $X_t = Y_k$ for $t \in [T_k, T_{k+1}).$

$(X_t)_{t \geq 0}$ is a cont. markov chain with generator Q

(b) Let $S = \mathbb{Z}_{\geq 0}^{\text{IN}} :$

~~With population n: it can go to n+1 or n-1 at a rate S_n~~

If population = n: It can either go to n+1 or n-1

cn+1): There are n processes leading to +1 ~~= n+1~~

$\text{Min } Z = \text{Min} \{ \xi_1, \dots, \xi_n \}, \xi_i \sim \text{Exp}(\lambda)$

$$\mathbb{P}(Z \geq t) = \mathbb{P}(\xi \geq t)^n = e^{-\lambda n t} \Rightarrow Z \sim \text{Exp}(n \cdot \lambda)$$

$\therefore n \rightarrow n+1$ at rate $n \cdot \lambda$

cn-1): In time $h, \mathbb{P}(n+1 \rightarrow n) = S_n \cdot h$

$\therefore n+1 \rightarrow n$ at a rate of $S_n.$

Q-matrix : $Q(n, n+1) = \lambda n$

$Q(n, n-1) = S_n$

$Q(n, n) = -(\lambda n + S_n)$

$Q(n, m) = 0$

Invariant Distribution:

(Birth - death cycle): π is invariant iff detailed balance holds

$$\pi_n \cdot (\lambda \cdot n) = \pi_{n+1} (\delta_{n+1}) \Rightarrow \pi_{n+1} = \pi_n (\lambda n) / \delta_{n+1}$$

$$= \lambda^n \cdot n! / \delta_{n+1} \dots \delta_2$$

\therefore Invariant distribution exists iff $\sum_{n \geq 1} \frac{\lambda^n n!}{\delta_{n+1} \dots \prod_{k=2}^n \delta_k} < \infty$.

If $\delta_n = \mu(n-1)$: $\pi_{n+1} = \pi_n \cdot (\lambda n) / \mu(n) = (\lambda/\mu) \pi_n$

$$\therefore \pi_{n+1} = \pi_1 \left(\frac{\lambda}{\mu}\right)^n$$

$$\sum_{n \geq 1} \pi_n = \pi_1 \left(\frac{\lambda}{\mu}\right) \sum_{n \geq 0} \left(\frac{\lambda}{\mu}\right)^n$$

Invariant distribution exists iff $\lambda < \mu$.

If $\lambda < \mu$: $\pi_n = (1 - \lambda/\mu) (\lambda/\mu)^{n-1}$

3) cas Let $(\varepsilon_n)_{n \geq 1}$ be iid $\text{Exp}(\lambda)$ R.V.:

$$T_0 = 0, \quad T_n = \sum_{k=1}^n \varepsilon_k; \quad (X_t)_{t \geq 0} \text{ defined by: } X_t = k \text{ for } t \in [T_k, T_{k+1})$$

$$N_t \sim \text{Poi}(\lambda t)$$

$$T_1 = \varepsilon_1, \quad T_{k+1} - T_k \sim \varepsilon_{k+1}$$

$$\therefore (T_1, T_2 - T_1, \dots, T_n - T_{n-1}) = (\varepsilon_1, \dots, \varepsilon_n)$$

$$f_{T_1, T_2 - T_1, \dots, T_n - T_{n-1}}(y_1, \dots, y_n) = \lambda^n e^{-\lambda(y_1 + \dots + y_n)} \quad (\text{product of independent R.V.})$$

$$\Lambda(T_1, \dots, T_n)^T = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ & & 1 \end{pmatrix} \begin{pmatrix} T_1 \\ T_2 - T_1 \\ \vdots \\ T_n - T_{n-1} \end{pmatrix}$$

$$\therefore f_{T_1, \dots, T_n}(y_1, y_1 + y_2, \dots, y_1 + \dots + y_n) = \left| \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ & & 1 \end{pmatrix} \right| \cdot f_{T_1, T_2 - T_1, \dots, T_n - T_{n-1}}(y_1, \dots, y_n)$$

$$= \lambda^n e^{-\lambda(y_1 + \dots + y_n)} \quad \text{Where } y_i \geq 0.$$

$$P(N_t = n) = \int_{A_1} \dots \int_{A_n} \lambda^n e^{-\lambda(x_1 + \dots + x_n)} \mathbb{1}_{0 \leq x_1 \leq \dots \leq x_n \leq t} dx_1 \dots dx_n$$

$$= \int_{A_1} \dots \int_{A_n} \lambda^n e^{-\lambda x_n} \mathbb{1}_{0 \leq x_1 \leq \dots \leq x_n \leq t} dx_1 \dots dx_n$$

$$= \int_{A_1} \dots \int_{A_n} \lambda^n e^{-\lambda x_n} \mathbb{1}_{0 \leq x_1 \leq \dots \leq x_n \leq t} dx_1 \dots dx_n$$

$$P(T_1 \in A_1, \dots, T_n \in A_n, N_t = n) = P(T_1 \in A_1, \dots, T_n \in A_n, T_n \leq t < T_{n+1})$$

$$= \int_{A_1} \dots \int_{A_n} \int_t^\infty \lambda^{n+1} e^{-\lambda x_{n+1}} dx_{n+1} \dots dx_1$$

$$= \int_{A_1} \dots \int_{A_n} \lambda^n e^{-\lambda t} \mathbb{1}_{0 \leq x_1 \leq \dots \leq x_n \leq t} dx_1 \dots dx_n$$

$$IP(T_1 \in A_1, \dots, T_n \in A_n \mid \{N_t = n\}) =$$

$$\int_{A_1 \dots} \int_{A_n} \lambda^n e^{-\lambda t} dt \quad \Bigg/ \quad e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

$$= \int_{A_1 \dots} \int_{A_n} \underbrace{1_{0 \leq x_1 \dots \leq x_n \leq t}}_{\text{Density of order statistics of } n \text{ iid } U[0, t] \text{ R.V.}} \frac{n!}{t^n} dt$$

Density of order statistics of n iid

$U[0, t]$ R.V.

$$\therefore \text{Joint Density: } f_{T_1, \dots, T_n}(x_1, \dots, x_n) \mid \{N_t = n\} = 1_{0 \leq x_1 \dots \leq x_n} \frac{n!}{t^n}$$

$$(b) \quad E \left[e^{\theta \sum_{j=1}^{N(t)} g(T_j, X_j)} \right] = E \left[E \left[e^{\theta \sum_{j=1}^n g(T_j, X_j)} \mid N(t) = n \right] \right]$$

$(T_1, \dots, T_n) \mid \{N_t = n\} \sim$ order statistic on U_1, \dots, U_n , (U_i are iid $U[0, t]$ R.V.)

$$\therefore E \left[e^{\theta \sum_{j=1}^n g(T_j, X_j)} \mid N_t = n \right] = E \left[e^{\theta \sum_{j=1}^n g(U_{(j)}, X_j)} \right],$$

θ is the ordering of (U_1, \dots, U_n)

$$= E \left[e^{\theta \sum_{j=1}^n g(\frac{U_j}{t}, X_{\theta^{-1}(j)})} \right] = E \left[e^{\theta \sum_{j=1}^n g(\frac{U_j}{t}, X_j)} \right]$$

as (X_1, \dots, X_n) are iid.

$$= E \left[e^{\theta g(\frac{U}{t}, X)} \right]^n$$

$$E \left[e^{\theta g(\frac{U}{t}, X)} \right] = \frac{1}{t} \int_0^t E \left[e^{\theta g(\frac{y}{t}, Y)} \right] dy = \Lambda$$

$$\therefore E \left[e^{\theta \sum_{j=1}^{N(t)} g(T_j, X_j)} \right] = E \left[\Lambda^{N(t)} \right] = e^{-\lambda t(1-\Lambda)} = e^{\lambda(\Lambda-1)t}$$

$$= e^{\lambda \int_0^t \frac{1}{t} E \left[g(\frac{y}{t}, X) \right] dy}$$

Let S be # of students at time T

$$S = \sum_{j=1}^{N(T)} 1_{T - T_j \leq X_j}$$

$$\therefore \mathbb{E}[e^{\theta \cdot S}] = \mathbb{E}[e^{\lambda \int_0^t \mathbb{E}[e^{\theta 1_{T-y \leq X}}] - 1} dy]$$

$$\mathbb{E}[e^{\theta 1_{T-y \leq X}}] = 1 + (e^{\theta} - 1) \mathbb{P}(X \geq T-y)$$

$$\therefore \mathbb{E}[e^{\theta \cdot S}] = \mathbb{E}[e^{\lambda \int_0^t \mathbb{P}(X \geq T-y) dy} (e^{\theta} - 1)]$$

If $Z \sim \text{Poi}(\mu)$, $\mathbb{E}[e^{\theta Z}] = \sum_{k=0}^{\infty} \frac{e^{-\mu} \mu^k}{k!} e^{\theta k} = e^{-\mu} e^{\mu e^{\theta}} = e^{\mu(e^{\theta} - 1)}$

$$\therefore S \sim \text{Poi}\left(\lambda \int_0^t \mathbb{P}(X \geq T-y) dy\right) \Rightarrow \text{Poisson R.V.}$$

(Pick T to represent a time interval of 9 am \rightarrow 5 pm)

4.) (a) $M(\lambda) / M(\mu) / 1$ Queue:

Consider Jump chain: Random Walk on $\mathbb{Z}_{\geq 0}$, $\mathbb{P}(+1) = \frac{\lambda}{\lambda + \mu}$, $\mathbb{P}(-1) = \frac{\mu}{\lambda + \mu}$.
 \therefore Transient iff $\frac{\lambda}{\lambda + \mu} > \frac{1}{2}$ iff $\lambda > \mu$.

Since jump chain recurrent iff $(X_t)_{t \geq 0}$ is recurrent:

$\lambda > \mu$: Transient

$\lambda < \mu$: Recurrent

If $\lambda < \mu$: Detailed Balance Equation $\pi_n \cdot \lambda = \mu \pi_{n+1}$
 $\therefore \pi_n = \left(\frac{\lambda}{\mu}\right)^n \cdot \pi_0$ (Invariant Measure)

\therefore Invariant distribution iff $\lambda/\mu < 1$; Recurrent \Rightarrow Non-explosive;

$\therefore \lambda = \mu$: null-recurrent

$\lambda < \mu$: +ve recurrent; Invariant Dist. $\pi_n = (1 - \lambda/\mu) (\lambda/\mu)^n$

* $M(\lambda) / M(\mu) / 1$ Definition: State Space = $\mathbb{Z}_{\geq 0}$

$$Q \text{ matrix: } Q(n, n+1) = \lambda \quad (n \geq 0)$$

$$Q(n, n-1) = \mu \quad (n \geq 1)$$

$$Q(n, n) = \lambda + 1_{n=0} \mu \quad (n \geq 0)$$

$$Q(n, m) = 0 \quad (\text{otherwise})$$

$M/M/\infty$ Queue: State Space $\mathbb{Z}_{\geq 0}$, Q matrix:

$$Q(n, n+1) = \lambda \quad (n \geq 0)$$

$$Q(n, n-1) = n \cdot \mu \quad (n \geq 1)$$

$$Q(n, n) = \lambda + n \mu \quad (n \geq 0)$$

$$Q(n, m) = 0 \quad (\text{otherwise})$$

$$\pi_n = (\lambda/\mu)^n / n! : \frac{\pi_n Q(n+1, n)}{\pi_{n+1} Q(n, n+1)} = \frac{n+1}{(\lambda/\mu)} \cdot \frac{\lambda}{(n+1)\mu} = 1$$

$\therefore \pi$ is in detailed balance, \Rightarrow Invariant

$$\sum_{n \geq 0} \pi_n = e^{(\lambda/\mu)} < \infty \Rightarrow \text{Invariant } (\pi_n / e^{\lambda/\mu})_n \text{ is an invariant distribution}$$

κ^* We know: If π is $(X_t)_{t \geq 0}$ invariant, $(\pi_n \cdot q_n)_{n \geq 0}$ is jump chain invariant

$$\text{Let } \tilde{\pi}_n = (\lambda/\mu)^n \cdot (\lambda + n\mu) / n! ; \tilde{\pi} \text{ is jump chain invariant}$$

$$\begin{aligned} \sum_{n \geq 0} \tilde{\pi}_n &= \lambda \sum_{n \geq 0} \frac{(\lambda/\mu)^n}{n!} + \lambda \sum_{n \geq 1} \frac{(\lambda/\mu)^{n-1}}{(n-1)!} \\ &= 2\lambda e^{(\lambda/\mu)} < \infty \end{aligned}$$

Since Invariant distribution exists for jump chain, jump chain is +ve recurrent \Rightarrow N Recurrent $\Rightarrow (X_t)_{t \geq 0}$ is recurrent \Rightarrow Non-explosive

Invariant distribution of $(X_t)_{t \geq 0}$ exists, Non-explosive \Rightarrow +ve recurrent.

$$\pi_n = e^{-(\lambda/\mu)} \frac{(\lambda/\mu)^n}{n!}$$

(ii) Let X_t be the length of taxi in queue, at time t :

Arrival rate = 3;

Service rate = Arrival rate of customers = 3 (Single Server)

$\therefore M(2) / M(3) / 1$ Queue, (+ve recurrent).

$\pi_n = \frac{1}{3} \left(\frac{2}{3}\right)^n$ is invariant distribution

In the long run: $IP(X_t = 0) \rightarrow \pi_0 = \frac{1}{3}$

$\therefore IP(\text{customer leaves}) \rightarrow \frac{1}{3}$

$\therefore IP(\text{arriving customer gets a taxi}) = \frac{2}{3}$

$$\begin{aligned} \text{Average \# of taxis: } \sum_{n=0}^{\infty} \left(\frac{1}{3}\right) \left(\frac{2}{3}\right)^n \cdot n &= \left(\frac{2}{3}\right) \sum_{n=1}^{\infty} \left(\frac{1}{3}\right) \left(\frac{2}{3}\right)^{n-1} n \\ &= \left(\frac{2}{3}\right) \left[\frac{1}{\left(\frac{1}{3}\right)^2} - 1 \right] = \frac{2}{3} \cdot \frac{1}{\left(\frac{1}{3}\right)^2} - \frac{2}{3} \\ &= -1 + \frac{1}{\left(\frac{1}{3}\right)} = 2. \end{aligned}$$

$$IE[\text{Time between two 0 taxis}] = \frac{1}{\pi_0 \cdot q_0} = \frac{1}{\frac{1}{3} \cdot 2} = \frac{3}{2}$$

$$IE[\text{Time to 1st taxi arrival from 0 queue length}] = \frac{1}{2}$$

$$\therefore IE[\text{Time from taxi arrival} \rightarrow \text{break}] = 1$$

* 2015 Q4 Part 1

4.) (a) Consider jump chain:

certain condition on $\{T_A = \infty\}$: that \tilde{q} is new a

$$\tilde{P}_{x,y} = \mathbb{P}(Y_1 = y)$$

$$\mathbb{P}(Y_{n+1} = y | Y_n = x, T_A) = \tilde{P}_{x,y}$$

$$\mathbb{P}(Y_0 = y_0, \dots, Y_n = y_n; T_A = \infty)$$

$$= P_{y_0, y_1, \dots, y_n} \cdot \mathbb{P}(T_A = \infty | Y_0 = y_0)$$

$$\therefore \mathbb{P}(Y_0 = y_0, \dots, Y_n = y_n | T_A = \infty) = \prod_{k=1}^n \tilde{P}_{y_{k-1}, y_k} \cdot \frac{h_{y_k}}{h_{y_{k-1}}}$$

$$\mathbb{E}_x[\tau_y] = h_x \cdot \mathbb{E}_x[\tau_y | T_A < \infty] + (1 - h_x) \mathbb{E}_x[\tau_y | T_A = \infty]$$

$$\mathbb{E}_x[\tau_y | T_A = \infty] = \sum_{k \geq 0} \mathbb{E}[\mathbb{P}(Y_k = y | T_A = \infty)]$$

$$= \frac{(1 - h_y)}{(1 - h_x)} \mathbb{E}[\tau_y]$$

$$\therefore \mathbb{E}_x[\tau_y] = h_x \mathbb{E}[\tau_y | T_A < \infty] + (1 - h_y) \mathbb{E}[\tau_y]$$

Applied Probability 2016

1.) (a) Let S be state space, $q(x) = -Q_{xx}$ ($x \in S$), $P(x, y) = \frac{1_{x \neq y} Q(x, y)}{q(x)}$

$(Y_n)_{n \geq 0}$ be jump chain: $Y_0 = X_0$, q & P is transition probability.

Let $(E_n)_{n \geq 1}$ be iid $\text{Exp}(1)$ R.V.:

$$J_0 = 0, \quad J_n = \sum_{k=1}^n E_k / q(Y_{k-1}) : X_t = Y_k \text{ for } t \in [J_k, J_{k+1})$$

(b) If state x is $(Y_n)_{n \geq 0}$ transient: $\mathbb{P}(\{Y_n = x\}_{n \geq 0}) < 1$
 x is non-absorbing;

Let $T = \sup \{n : Y_n = x\}$; $T < \infty$ A.S. (x is Transient)

\therefore If $t > J_{T+1}$, $X_t \neq x$

$\therefore \{t : X_t = x\} \subseteq [0, J_{T+1}] \Rightarrow$ ~~Bounded A.S.~~

$\therefore \{X_t\}_{t \geq 0}$ transient

Since $J_{T+1} < \infty$ ~~A.S.~~ if $T < \infty$ A.S. : $\{t : X_t = x\}$ is bounded A.S.

$\therefore X$ is $(X_t)_{t \geq 0}$ transient.

(c) Let $S = \mathbb{Z}$, $Q(n, n \pm 1) = \frac{(1 + |n|)^2}{2}$, $Q(n, n) = (1 + |n|)^2$.

Discrete chain: $(Y_n)_{n \geq 0}$ is a symmetric random walk on $\mathbb{Z} \Rightarrow$ Null-recurrent.

$\therefore (X_t)_{t \geq 0}$ is recurrent; Consider $\pi_n = \frac{1}{(1 + |n|)^2}$:

$$\pi_n Q(n, n \pm 1) = \pi_{n \pm 1} Q(n \pm 1, n) = \frac{1}{2} \Rightarrow \pi \text{ is in detailed balance}$$

\therefore Invariant measure.

$$\sum_{n \in \mathbb{Z}} \frac{1}{(1 + |n|)^2} < \infty \Rightarrow \text{Invariant distribution} \Rightarrow \{X_t\} \text{ recurrent}$$

(~~is~~ recurrent \Rightarrow not explosive)

\therefore (Counterexample)

- (d) Consider jump chain: Random Walk on \mathbb{Z} , $1P(+1) = 2/3$, $1P(-1) = 1/3$
 \therefore Jump chain is transient $\Rightarrow (X_n)_{n \geq 0}$ is transient.

Solve: $\pi_n (2 \cdot 3^{1n}) = \pi_{n+1} \cdot 3^{1n+1}$

($n \geq 0$) $\therefore \pi_{n+1} = \frac{2 \cdot 3^n}{3^{n+1}} \pi_n = \left(\frac{2}{3}\right) \pi_n$

$\therefore \pi_n = \left(\frac{2}{3}\right)^n \pi_0$

($n < 0$): $\pi_n = \pi_{n+1} \cdot \frac{3^{-(n+1)}}{2 \cdot 3^{-n}} = \frac{\pi_{n+1}}{6}$

$\therefore \sum_{n \in \mathbb{Z}} \pi_n < \infty \Rightarrow$ Invariant Distribution exists

- (e) Let $h_n = \mathbb{E}_n [T_0]$:

$$h_n = \frac{1}{3} h_{n+1} + \frac{2}{3} h_{n+1} + \frac{1}{3} h_{n-1}$$

$$\therefore h_n = h_{n+1}$$

$$\therefore h_n = A + B\left(\frac{1}{3}\right)^n + C$$

If $1P(T_0 = \infty) > 0$: Since chain is recurrent, if $\mathbb{E}_0 [T_0] < \infty$,
 0 is +ve recurrent;

Since chain is irreducible: All states are +ve-recurrent \Rightarrow Recurrent
 \therefore Contradiction

- (e) Jump chain: Symmetric Random Walk on \mathbb{Z}^d
 \therefore Recurrent iff $d \leq 2$.

If $d \leq 2$: $\pi_x = \frac{1}{q(x)}$ & solves detailed balance

\therefore +ve recurrent iff $\sum_{x \in \mathbb{Z}^d} \frac{1}{\min_{y \sim x} \{1, \|x\|^d\}} < \infty$. (*)

Since $\|\cdot\|_1$ is equivalent to $\|\cdot\|_\infty$, (*) is equivalent to

$$\sum_{i=1}^N 1_{n_i > 0} \pi_{n-e_i} \sum_{j \neq i} \pi_{n+e_j-e_i} \mu_j P_{j,i}^{(n_j+1)} =$$

$$\sum_{i=1}^N 1_{n_i > 0} \pi_n \left\{ \sum_{j \neq i} \frac{(\lambda_j / n_{j+1})}{(\lambda_i / n_i)} \mu_j P_{j,i} \right\}$$

$$= \sum_{i=1}^N 1_{n_i > 0} \pi_n \frac{\mu_i}{\lambda_i} \sum_{j \neq i} \lambda_j P_{j,i}$$

$$\sum_{i=1}^N 1_{n_i > 0} \pi_n \sum_{j \neq i} \frac{(\lambda_j / n_{j+1})}{(\lambda_i / n_i)} \mu_j P_{j,i}^{(n_j+1)}$$

$$= \sum_{i=1}^N 1_{n_i > 0} \pi_n \frac{n_i}{\lambda_i} \sum_{j \neq i} \lambda_j \mu_j P_{j,i} = \sum_{i=1}^N 1_{n_i > 0} \pi_n \frac{n_i \mu_i}{\lambda_i} \lambda_i$$

$$= \mu_i \lambda_i$$

$$= \left(\sum_{i=1}^N 1_{n_i > 0} \mu_i n_i \right) \pi_n \Rightarrow \text{Solves Invariant Equation.}$$

cs) Not independent: Let $A^{(k)}$ be length of queue k

$$IP(A^{(k)}(1) = k) > 0$$

$$\text{But } \prod_{k=1}^N IP(A^{(k)}(1) = k) > 0, \quad \sum_{k=1}^N IP(A^{(k)}(1) = k) = 0 \quad (1 \leq k \leq N)$$

\therefore Not independent

$$(M_t)_{t \geq 0}$$

3.) ca) Let $(N_t)_{t \geq 0}$ be a independent Poisson processes of rate λ, μ respectively:

Superposition: $(N_t + M_t)_{t \geq 0}$ is a Poisson process of rate $\lambda + \mu$

Thinning: Let $(\alpha_n)_{n \geq 1}$ be independent of $(N_t)_{t \geq 0}$ iid $Be(p)$ r.v.

Let \dots be points in N_t

Let Y be process in $\mathbb{N} \times \mathbb{Z}_{\geq 0}$, with jump at time t iff N jumps at t , $Z_{X_t} = 1$.

\exists Y is a Poisson rate $p\lambda$ process, $Z = X - Y$ is a Poisson rate $(1-p)\lambda$ process; Z, Y independent.

Proof of Superposition: $(N_t)_t, (M_t)_t$ have independent increments $\Rightarrow (N_t + M_t)_{t \geq 0}$ has independent increments

$$\begin{aligned} \mathbb{P}(N_{t+h} + M_{t+h} = N_t + M_t) &= \mathbb{P}(N_{t+h} = N_t, M_{t+h} = M_t) \\ &= (1 - \lambda h + o(h)) (1 - \mu h + o(h)) \\ &= 1 - (\lambda + \mu)h + o(h) \quad (\text{Uniformly in } h) \end{aligned}$$

$$\begin{aligned} \mathbb{P}(N_{t+h} + M_{t+h} - N_t - M_t = \frac{1}{2}) &= \mathbb{P}(N_{t+h} - N_t = 1, M_{t+h} - M_t = 0) + \\ &\quad \mathbb{P}(N_{t+h} - N_t = 0, M_{t+h} - M_t = 1) + o(h) \\ &= e^{-\lambda}(\lambda h + o(h)) e^{-\mu} + e^{-\lambda} e^{-\mu}(\mu h + o(h)) + o(h) \\ &= (\lambda + \mu)h + o(h) \end{aligned}$$

(Uniformly in all t)

\therefore (Infinitesimal Definition): $(N_t + M_t)_{t \geq 0}$ is a Poisson process of rate $\lambda + \mu$.

(b) Let $(X_t)_t, (Y_t)_t$ be independent Poisson processes of rate λ .

Define: $N_t = X_t \quad (t \geq 0)$
 $N_t = Y_{-t} \quad (t < 0)$

N_t, N_s : If $t, s \geq 0$: have the same sign
 ≤ 0 :
 $s \leq 0 \leq t$

$N_t - N_s = X_t - X_s \sim \text{Poi}(\lambda(t-s))$
 $N_t - N_s = Y_{-s} - Y_{-t} \sim \text{Poi}(\lambda(t-s))$
 $N_t - N_s = X_t - Y_s$

$$\text{Let } N_t = X_t \quad (t \geq 0) \\ = -Y_{-t} \quad (t \leq 0)$$

$$\text{If } 0 \leq s \leq t: N_t - N_s = X_t - X_s \sim X_{t-s} \quad (\text{indp, stat. increment}) \\ \sim \text{Poi}(\lambda(t-s))$$

$$0 \geq t \geq s: N_t - N_s = -\cancel{X_{-t}} + \cancel{Y_{-s}} \sim Y_{-(s-t)} \sim \text{Poi}(\lambda(t-s))$$

$$t \geq 0 \geq s: N_t - N_s = N_t - N_0 + N_0 - N_s = X_t + \cancel{Y_{-s}} \\ \sim \text{Poi}(\lambda(t-s)) \quad (\text{Superposition})$$

($N_0 = 0$ by definition)

Any increment is independent, stationary as $(X_t)_{t \geq 0}, (Y_t)_{t \geq 0}$ have independent, stationary increments

(Independent Increments): $t_0 < t_1 \dots < t_k = 0 < t_{k+1} \dots < t_N$

$$(N_{t_1} - N_{t_0}, \dots, N_{t_N} - N_{t_{N-1}}) = (Y_{-t_0} - Y_{-t_1}, \dots, Y_{-t_{k-1}} - Y_0; X_{t_{k+1}} - X_0, \dots, X_{t_N} - X_{t_{N-1}}) \Rightarrow \text{Independent}$$

Since $N_t - N_s \sim \text{Poi}(\lambda(t-s))$ is only dependent on $t-s$; increments are stationary.

\therefore Process Exists

cc) Let \tilde{N} be shifted process:

$$\tilde{N}_t - \tilde{N}_s = N_{t-c} - N_{s-c} \sim \text{Poi}(\lambda(t-s))$$

& increments remain stationary, independent as

$$(\cancel{N_{t_1-t_0}}, (\tilde{N}_{t_1} - \tilde{N}_{t_0}, \dots, \tilde{N}_{t_N} - \tilde{N}_{t_{N-1}}) = (N_{t_1-c} - N_{t_0-c}, \dots, N_{t_N-c} - N_{t_{N-1}-c})$$

\Rightarrow Independent, stationary increments

$\therefore \tilde{N}$ is the same process

(1) Thinning on each point: ($IP = 1/2$)

$N \rightarrow X, N-X$; $X, N-X$ are independent $Poi(\lambda/2)$ processes

Shift X by $+1$, $N-X$ by -1 : ^{Remains} Independent $Poi(\lambda/2)$ processes

Combine: $Poi(\lambda)$ process. \Rightarrow Remains a $Poi(\lambda)$ process

4.) (a) Let $(\xi_n)_{n \geq 1}$ be iid, ≥ 0 R.V., $E[\xi_1] > 0$.

$N_t = \max \{k \geq 0 : \xi_1 + \dots + \xi_k \leq t\}$ is a renewal process

$$\sum_{k=1}^{N_t} \xi_k \leq t \leq \sum_{k=1}^{N_t+1} \xi_k$$

Claim: $N_t \rightarrow \infty$ A.S. \forall

$$\exists \varepsilon > 0 \text{ s.t. } IP(\xi_1 > \varepsilon) > 0 \Rightarrow \sum_{n \geq 1} IP(\xi_n > \varepsilon) = \infty$$

$$\stackrel{2^{nd} \text{ Borel Cantelli}}{\Rightarrow} \xi_n > \varepsilon \text{ i.o.} \stackrel{A.S.}{\Rightarrow} N_t \rightarrow \infty \text{ A.S.}$$

$$\dots \text{ (Strong law)} \quad \frac{1}{N(t)} \sum_{k=1}^{N_t} \xi_k \rightarrow \lambda^{-1} \text{ A.S.}$$

$$\frac{N(t)+1}{N(t)} \rightarrow 1 \text{ (A.S.)} \Rightarrow \frac{1}{N(t)} \sum_{k=1}^{N_t+1} \xi_k \rightarrow \lambda^{-1} \text{ A.S.}$$

$$\therefore t/N(t) \rightarrow 1/\lambda \text{ A.S.} \Rightarrow \frac{N(t)}{t} \rightarrow \lambda \text{ A.S.}$$