

STA447 Final Review

Markov Chains

- A (discrete time, discrete space, time homogeneous) Markov chain is specified by 3 ingredients.
 - ① A state space S , non empty finite or countable set.
 - ② Transition probabilities $\{P_{ij}\}_{i,j \in S}$
 - ③ Initial probabilities $\{\nu_i\}_{i \in S}$

$$\nu_i^{(n)} = P(X_n=i), \quad \nu_i^{(0)} = \nu_i, \quad \nu_j^{(0)} = P(X_0=j) = \sum_{i \in S} P(X_0=i, X_1=j) = \sum_{i \in S} \nu_i p_{ij}$$

Law of Total Probability

• "n-step transition": $p_{ij}^{(n)} = P(X_{n+1}=j | X_0=i)$

• Chapman-Kolmogorov equation: $p_{ij}^{(n+1)} = \sum_k p_{ik}^{(n)} p_{kj}^{(n)}$

Classification of States

- a state i is recurrent if $P_i(X_n=i \text{ for some } n \geq 1) = 1$
otherwise, transient.

- RECURRENT THEOREM:** i recurrent iff $\sum_{n=1}^{\infty} p_{ii}^{(n)} = \infty$ iff $P_i(N_{(i)} = \infty) = 1$
 i transient iff $\sum_{n=1}^{\infty} p_{ii}^{(n)} < \infty$ iff $P_i(N_{(i)} = \infty) = 0$

To prove this, let $f_{ij} = P_i(X_n=j \text{ for some } n \geq 1)$

Then iff recurrent $f_{ii} = 1$, if transient $f_{ii} < 1$

Communicating States

i communicates j : $f_{ij} > 0$.

A MC is irreducible if $\forall i, j \in S, i \rightarrow j$.

CASE THEOREM: For an irreducible MC either ① $\sum_{n=1}^{\infty} p_{ij}^{(n)} = \infty \forall i, j \in S$, all states recurrent; ② $\sum_{n=1}^{\infty} p_{ij}^{(n)} < \infty \forall i, j \in S$, all states transient.

SUM LEMMA: if $i \rightarrow k$, $k \rightarrow j$ & $\sum_{l=1}^{\infty} p_{kl}^{(n)} = \infty$, then $\sum_{l=1}^{\infty} p_{lj}^{(n)} = \infty$

FINITE SPACE THM: an irreducible MC on a finite state space always falls into case (a), i.e. $\sum_{n=1}^{\infty} p_{ij}^{(n)} = \infty \forall i, j \in S$ and all states are recurrent.

• $T_i = \min \{n \geq 1 : X_n = i\}$ ($T_i = \infty$ if never hit i)

HIT LEMMA: If $j \rightarrow i$ with $j \neq i$. Then $P_j(T_i < T_j) > 0$.

F-LEMMA: If $j \rightarrow i$ & $f_{ij} = 1$, then $f_{jj} = 1$

STRONGER-RECURRENT THM: If chain irreducible: TFRE

① $k \in S$ with $\sum_{n=1}^{\infty} p_{kk}^{(n)} = \infty$

② $\forall i, j \in S, \sum_{n=1}^{\infty} p_{ij}^{(n)} = \infty$

③ $\exists k \in S, f_{kk} = 1$

④ $\forall j \in S, f_{jj} = 1$

⑤ $\forall i, j \in S, f_{ij} = 1$

Stationary Distribution

- Let π_i be a probability distribution on S , i.e. $\pi_i \geq 0 \forall i \in S$, $\sum_{i \in S} \pi_i = 1$.
- ~~π_i~~ π_i is a stationary for a Markov chain $P(P_{ij})$ if $\sum_{i \in S} \pi_i P_{ij} = \pi_j \forall j \in S$.
- a Markov chain is reversible (or time reversible) wrt a prob dist'n $\{\pi_i\}$ if $\pi_i P_{ij} = \pi_j P_{ji} \forall i, j \in S$.
- PROPOSITION.** if chain is ~~irreducible~~ reversible wrt $\{\pi_i\}$ then $\{\pi_i\}$ is a stationary distribution. (Converse is false).

Obstacles to Convergence

- The period of a state i is gcd of the set $\{n \geq 1; P_{ii}^{(n)} > 0\}$
- If period of each state is 1, chain is aperiodic.
- if $P_{ii} > 0$, then period of i is 1.
- Simple random walk / Ehrenfest's Chain period 2.
- ~~EQUAL PERIODS LEMMA~~: if $i \leftrightarrow j$ then the periods of i & j are equal.

COR: if chain irreducible then all states the same period.

COR: if chain irreducible & $P_{ii} > 0$ for some state i , then the chain is aperiodic.

MARKOV CHAIN CONVERGENCE THEOREM

If a MC is irreducible and aperiodic and has a stationary distribution $\{\pi_j\}$ then $\lim_{n \rightarrow \infty} P_{ij}^{(n)} = \pi_j$ for $\forall i, j \in S$ and $\lim_{n \rightarrow \infty} P(X_n = j) = \pi_j$ for any initial prob $\{x_j\}$

STATIONARY RECURRENT LEMMA. If chain irreducible, and has stationary dist, then it is recurrent.

COR: If chain is irreducible and aperiodic, then at most one stationary dist.

PERIODIC CONVERGENCE THM. Sps chain irreducible, period $b \geq 2$, stat dist $\{\pi_i\}$. $\forall i, j \in S$, $\lim_{n \rightarrow \infty} \frac{1}{b} [P_{ij}^{(nb)} + \dots + P_{ij}^{(nb+1)}] = \pi_j$ also

$$\lim_{n \rightarrow \infty} \frac{1}{b} P[X_n = j \text{ or } X_{n+1} = j \text{ or } \dots \text{ or } X_{n+b-1} = j] = \pi_j$$

COR: If MC irreducible (not necessarily aperiodic), at most one stat dist.

MCMC

π_i

$S = \mathbb{Z}$, $\{\pi_i\}$ be any prob dist on S . Assume $\pi_i > 0 \forall i$.

$$\left. \begin{array}{l} \text{Let } p_{i,i+1} = \frac{1}{2} \min \left[1, \frac{\pi_{i+1}}{\pi_i} \right] \\ p_{i,i-1} = \frac{1}{2} \min \left[1, \frac{\pi_{i-1}}{\pi_i} \right] \\ p_{i,i} = 1 - p_{i,i+1} - p_{i,i-1} \\ p_{ij} = 0 \text{ o.w.} \end{array} \right\} \text{ s.t. } \lim_{n \rightarrow \infty} P_{ij}^{(n)} = \pi_j$$

- Equivalent algorithmic version: Given X_{n-1} , let Y_n equal $X_{n-1} \pm 1$ (prob $\frac{1}{2}$) each and $U_n \sim \text{Uniform}[0, 1]$ (indep.) and

$$X_n = \begin{cases} Y_n, & U_n < \frac{\pi_{Y_n}}{\pi_{X_{n-1}}} \\ X_{n-1}, & \text{o.w.} \end{cases}$$

Then $\pi_i p_{i,i+1} = \pi_i \cdot \frac{1}{2} \min[1, \frac{\pi_{i+1}}{\pi_i}] = \frac{1}{2} \min[\pi_i, \pi_{i+1}]$

$$\pi_{i+1} p_{i+1,i} = \pi_{i+1} \cdot \frac{1}{2} \min[1, \frac{\pi_i}{\pi_{i+1}}] = \frac{1}{2} \min[\pi_{i+1}, \pi_i]$$

So $\pi_i p_{ij} = \pi_j p_{ji}$ if $j = i+1 \forall j \in S$. reversible. $\{\pi_i\} \rightarrow \text{stationary}$ (easy part)

Random walks on graphs

$w(u,v)$ weight on edge (u,v) , $d(u) = \sum_{v \in V} w(u,v)$ (degree of vertex u)

$$p_{uv} = \frac{w(u,v)}{d(u)}$$

$$Z = \sum_{u \in V} d(u) = \sum_{u,v \in V} w(u,v) \quad (\text{unweighted case: } Z = 2 \times (\# \text{ of edges}))$$

$$\pi_u = \frac{d(u)}{Z}$$

If graph is connected, then chain is irreducible.

If graph is bipartite, then chain has period 2.

THM: for random walk on any connected graph with $Z < \infty$, $\lim_{n \rightarrow \infty} \frac{1}{2} [P_{uv}^{(n)} + P_{vu}^{(n)}] = -\frac{d(v)}{Z}$

Gambler's Ruin

$T_i = \inf\{n \geq 0 : X_n = i\}$ be first time having i dollars.

write $P_a(T_c < T_b)$ as $s(a)$

Gambler's formula: $s(a) = \begin{cases} \frac{(\frac{1-p}{p})^a - 1}{(\frac{1-p}{p})^c - 1}, & p \neq \frac{1}{2} \\ \frac{a}{c}, & p = \frac{1}{2} \end{cases}$

Martingales

A sequence $\{X_n\}_{n=0}^{\infty}$ of r.v. is a martingale if $E|X_n| < \infty$ for each n and also $E(X_{n+1} | X_0, \dots, X_n) = X_n$ (i.e. it stays same on average).

SPECIAL CASE: if $\{X_n\}$ is a MC (with $E|X_n| < \infty$) then $E(X_{n+1} | X_0, \dots, X_n)$

$$= \sum_j j P[X_{n+1} = j | X_0, \dots, X_n] = \sum_j j P_{X_n, j} \text{ so martingale if } \sum_j j P_{ij} = i \text{ for all } i$$

double-expectation formula

$$E(X_{n+1}) = E[E(X_{n+1} | X_0, \dots, X_n)] = E(X_n). \text{ i.e. } E(X_n) = E(X_0) \forall n.$$

A non-negative-integer-valued random variable T is a stopping time for $\{X_n\}$ if the event $\{T = n\}$ is determined by $\{X_0, X_1, \dots, X_n\}$.

"can't look into future before deciding to stop."

• OPTIONAL STOPPING LEMMA:

If $\{X_n\}$ martingale, with stopping time which is bdd ($\exists M < \infty$ with $P(T \leq M) = 1$), then $E(X_T) = E(X_0)$.

• OPTIONAL STOPPING THM:

If $\{X_n\}$ martingale, with stopping time T and $P(T < \infty) = 1$ and $E|X_T| < \infty$, and $\lim_{n \rightarrow \infty} E(X_n \mathbf{1}_{T > n}) = 0$, then $E(X_T) = E(X_0)$

• OPTIONAL STOPPING COR:

If $\{X_n\}$ is martingale with stopping time T , which is bdd up to time T " (i.e. $\exists M < \infty$ with $P(|X_n| \mathbf{1}_{n \leq T} \leq M) = 1$ for $\forall n$) and $P(T < \infty) = 1$, then $E(X_T) = E(X_0)$

• WALD'S THM: Sps $X_n = a + Z_1 + \dots + Z_n$ with $\{Z_i\}$ iid, finite mean m .

Let T be a stopping time for $\{X_n\}$ which has finite mean, $E(T) < \infty$, then $E(X_T) = a + mE(T)$

• special case, if $m=0$, $\{X_n\}$ is a martingale.

LEMMA: Let $X_n = a + Z_1 + \dots + Z_n$, $\{Z_i\}$ iid with mean 0, variance $\nu < \infty$.

Let $Y_n = (X_n - a)^2 - n\nu = (Z_1 + \dots + Z_n)^2 - n\nu$, then $\{Y_n\}$ is a martingale.

COR: If $\{X_n\}$ is Gambler's Ruin with $p = \frac{1}{2}$, and $T = \inf\{n \geq 0 : X_n = 0 \text{ or } c\}$ then $E(T) = \text{Var}(X_T) = a(c-a)$

~~All Abo~~

Martingale Convergence Thm

Any non-negative martingale converges w.p. 1 to some random variable X .

• non-negative, can't spread out forever, since it's a martingale; can't "drift" anywhere.

s.s.r.w. - martingale but not non-negative, does not converge

s.s.r.w. stopped at zero - converges.

s.r.w. with $p = \frac{2}{3}$ stopped at zero - non-negative but not converge b/c not a martingale.

Branching Process

Brownian Motion

$\{B_t\}_{t \geq 0}$ is the limit as $M \rightarrow \infty$ of $\{Y_t^{(M)}\}$

$$Y_t^{(M)} = \frac{1}{\sqrt{M}} (Z_1 + Z_2 + \dots + Z_M)$$

Thus $E(Y_t^{(M)}) = 0$, $\text{Var}(Y_t^{(M)}) = \frac{1}{M} (tM) = t$, as $M \rightarrow \infty$ by CLT, $Y_t^{(M)} \xrightarrow{D} N(0, t)$
so $B_t \sim N(0, t)$

Also $B_s - B_t \sim \text{Normal}(0, s-t)$ independent of B_t ($s \geq t > 0$)

$$\begin{aligned} \text{Cov}(B_t, B_s) &= E(B_t B_s) = E(B_t [B_s - B_t + B_t]) = E(B_t [B_s - B_t]) + E(B_t^2) \\ &= E(B_t) E(B_s - B_t) + E(B_t^2) \\ &= 0 \cdot 0 + t \end{aligned}$$

In general, $\text{Cov}(B_t, B_s) = \min(t, s) = t$ (covariance structure)

• Definition: Brownian motion is a process $\{B_t\}_{t \geq 0}$ satisfying the above property and with continuous sample paths (i.e. $t \rightarrow B_t$ is continuous)

$$\begin{aligned} \text{e.g. } E[(B_2 + B_3 + 1)^2] &= E[(B_2)^2] + E[(B_3)^2] + 1^2 + 2E[B_2 B_3] + 2E[B_2(1)] + 2E[B_3(1)] = 2 + 3 + 1 + 2(2) \\ \text{Var}(B_3 + B_5 + 7) &= E[(B_3 + B_5)^2] = E(B_3^2) + E(B_5^2) + 2E(B_3 B_5) = 3 + 5 + 2(3) = 10 \end{aligned}$$

$$\bullet E[Y_s | \{Y_r\}_{r \leq t}] = E[E[Y_s | \{B_r\}_{r \leq t}] | \{Y_r\}_{r \leq t}] = E[Y_t | \{Y_r\}_{r \leq t}] = Y_t \quad \text{law of iterated expectations}$$

week 8 p31.

Financial Modeling

• common model for stock price: $X_t = x_0 \exp(\mu t + \sigma B_t)$

"fair pair" Black-Scholes formula

$$X_0 \Phi\left(\frac{(r + \sigma^2/2)s - \log(K/x_0)}{\sigma \sqrt{s}}\right) - e^{-rs} K \Phi\left(\frac{(r - \sigma^2/2)s - \log(K/x_0)}{\sigma \sqrt{s}}\right)$$

RETURN TIME THM: If an irreducible MC on

Sequence Waiting Times

a discrete S has stat. dist π_i , $i \in S$, mean return

T be the first time the sequence "HTH" is completed. $E(T)$? time $m_i =$

Let X_n be the amount of the desired sequence (HTH) that the chain has "achieved" so far. If begins with HHTTHHT, then $X_0 = 0$, $X_1 = 1$, $X_2 = 1$, $X_3 = 2$, $X_4 = 0$,

So, $S = \{0, 1, 2, 3\}$ $X_0 = 0$, $X_T = 3$

$X_5 = 1$, $X_6 = 2$

• $P_{01} = P_{12} = P_{23} = \frac{1}{2}$ (prob of continuing the process)

• $P_{00} = P_{20} = \frac{1}{2}$. But ~~$P_{10} = P_{11} = \frac{1}{2}$~~ $P_{11} = \frac{1}{2}$

• $P_{31} = P_{20} = \frac{1}{2}$, o.w. 0.

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & 0 & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \end{pmatrix}$$

Then stat dist'n:

$$\begin{cases} \pi_0 = \pi_1 \cdot P_{01} + \pi_2 \cdot P_{20} + \pi_3 \cdot P_{30} + \pi_0 \cdot P_{00} \\ \pi_3 = \pi_1 \cdot P_{13} + \pi_0 \cdot P_{03} + \pi_2 \cdot P_{23} + \pi_3 \cdot P_{33} \\ \dots = \dots = \dots = \dots \\ \pi_1 = \frac{1}{\pi_0} = \frac{1}{0.1} = 10 \end{cases}$$

$$\pi = (0.8, 0.4, 0.2, 0.1)$$

dallen.

But the prob from $3 \rightarrow 3$ is the same as $0 \rightarrow 3$. ✓

• Another approach: (HTH) using martingale.

(Assume at time each n a better. bets 1 on H, then if win bet 2 on T, if win bet 4 on H. Stopped as soon as win 3 in a row or lose once)

S_n be total amount won by all betters by time n .

$\{S_n\}$ is a martingale with stopping time T .

$$S_T = -(T-3) + (-1)^T + 1^T$$

Poisson Process $\lambda = \text{average fires}$

$$P(\# \text{ fires} = k) = e^{-\lambda} \frac{\lambda^k}{k!}$$

$T_0 = 0, T_n = Y_1 + Y_2 + \dots + Y_n \forall n \geq 1$ (nth arrival time)

$N(t) = \max \{n \geq 0 : T_n \leq t\} = \#\{n \geq 1 : T_n \leq t\} = \# \text{ arrivals up to time } t$

- "counting process" e.g. $N(t) = \# \text{ fires between times } 0 \text{ & } t$.

"poisson process with intensity λ "

$$P(N(t)=k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, k=0,1,2,\dots$$

$$E(N(t)) = \lambda t, \text{Var}(N(t)) = \lambda t$$

• "memoryless" property of the $\text{Exp}(\lambda)$ dist'n: $\forall a, b > 0, P(Y_n > a+b | Y_n > a) = P(Y_n > b) = e^{-\lambda b}$

$$N(t+s) \sim N(s) \sim N(t) \sim \text{Poisson}(\lambda t)$$

if $0 \leq a < b \leq c < d$ then $N(d) - N(c)$ independent of $N(b) - N(a)$ independent

• more generally: $0 \leq t_1 \leq s_1 \leq t_2 \leq s_2 \leq \dots \leq t_k \leq s_k$ increments

$$N(s_i) - N(t_i) \sim \text{Poisson}(\lambda(s_i - t_i)) \text{ and } (N(s_i) - N(t_i))_{i=1}^k \text{ are all}$$

independent. (independent Poisson increments)

• A Poisson process with intensity $\lambda > 0$ is a collection $\{N(t)\}_{t \geq 0}$ of random non-decreasing integer counts $N(t)$ satisfying

(a). $N(0) = 0$.

(b). $N(t) \sim \text{Poisson}(\lambda t) \forall t \geq 0$

(c). independent Poisson increments.

$$0 < t < s, P(N(t) = 1 | N(s) = 1) = \frac{t}{s}$$

$$P(N(4) = 1 | N(5) = 3) = \frac{P(N(4) = 1, N(5) - N(4) = 2)}{P(N(5) = 3)}$$

C-time, MC,
large t , $P^{(t)} = ?$
Date

Sum of poisson is Poisson.

Continuous-Time, discrete-space MC

- Brownian motion & Poisson processes defined in continuous time.
- A ~~continuous-time~~ (time-homogeneous, non-explosive) Markov process, on a countable (discrete) state space S , is a collection $\{X(t)\}_{t \geq 0}$ of r.v.s.t.

$$P(X_0 = i_0, X_{t_1} = i_1, \dots, X_{t_n} = i_n) = \pi_{i_0}^{(t_1)} \pi_{i_1}^{(t_2-t_1)} \dots \pi_{i_{n-1}}^{(t_n-t_{n-1})}$$

- initial prob. & transition prob.

$$P^{(t)} = (P_{ij}^{(t)})_{i,j \in S} \text{ = matrix version}$$

$$P_{ij}^{(s+t)} = \sum_{k \in S} P_{ik}^{(s)} P_{kj}^{(t)} \quad P^{(s+t)} = P^{(s)} P^{(t)} \quad \text{Chapman-Kolmogorov Equation}$$

- process's generator $g_{ij} = \lim_{t \downarrow 0} \frac{P_{ij}^{(t)} - \delta_{ij}}{t} = P_{ij}'(0)$

$$G = P' = \lim_{t \downarrow 0} \frac{P^{(t)} - I}{t}$$

- if $t > 0$ is small, $G \approx \frac{P^{(t)} - I}{t}$ so $P^{(t)} \approx I + tG$ i.e. $P_{ij}^{(t)} \approx \delta_{ij} + t g_{ij}$

- reversibility still implies stationary. (or $\pi G = 0$) $\begin{matrix} = 1 & \text{if } i=j \\ 0 & \text{o.w.} \end{matrix}$

CONTINUOUS-TIME MARKOV CONVERGENCE THM.

if a continuous-time MC is irreducible. has a stationary-distribution π , then $\lim_{t \rightarrow \infty} P_{ij}^{(t)} = \pi_j$ for all $i, j \in S$.

(automatically aperiodic)

- Generator of Markov process $\{X_t\}$

Queueing Theory

T_n = time of arrival of n^{th} customer ($T_0 = 0$)

$T_n = T_n - T_{n-1}$ = inter-arrival time b/w $(n-1)^{\text{st}}$ and n^{th} customers.

S_n = time it takes to serve the n^{th} customer.

$Q(t)$ = number of customers in the system at time $t \geq 0$.

M/M/1 QUEUE:

$T_0, T_{n+1} \sim \text{Exp}(\lambda)$, $S_n \sim \text{Exp}(\mu)$, all indep., $\lambda, \mu > 0$.

(So $\{T_n\}$ are arrival times of a Poisson process with intensity λ)

Then $[Q(t)]$ is a Markov process.

General (G/G/1) Queue:

D_n = time of departure of the n th customer.

$W_n = \max(0, D_{n-1} - T_n)$ = the amount of time that the n^{th} customer has to wait. (With $W_0 = 0$)

LINDLEY'S COR: For $n \geq 1$, $W_n = \max(0, W_{n-1} + S_{n-1} - T_n)$

LINDLEY'S COR: $W_n = \max_{0 \leq k \leq n} \sum_{i=k+1}^n (S_{i-1} - T_i)$

THM: For a general (G/G/1)

(a). if $E(Y_n) < E(S_n)$ then $\lim_{n \rightarrow \infty} W_n = \infty$ w.p. 1.

(Hence, $\lim_{n \rightarrow \infty} W_n = \infty$ in prob, so $\forall M < \infty$, $\lim_{n \rightarrow \infty} P(W_n > M) = 1$).

(b). if $E(Y_n) > E(S_n)$, then $\{W_n\}$ is bdd in probability, i.e. for any $\varepsilon > 0 \exists M < \infty$ s.t. $P(W_n > M) < \varepsilon$ for all $n \in N$.

(c). if $E(Y_n) = E(S_n)$, and S_{n-1} & T_n are not both constant

(i.e. $\text{Var}(S_{n-1} - T_n) > 0$) then $W_n \rightarrow \infty$ in prob. (but not w.p. 1)

Quantum Mechanics

Renewal Theory:

arrival times $\{T_n\}$, $T_0 = 0$, $T_n = Y_1 + Y_2 + \dots + Y_n$ where $\{Y_n\}_{n=1}^\infty$ are independent interarrival times with $\{Y_n\}_{n=2}^\infty$ i.i.d.

$N(t) = \max\{n \geq 0; T_n \leq t\} = \#\{n \geq 1; T_n \leq t\}$ is a renewal process

If $\{Y_n\}_{n=1}^\infty$ all i.i.d then the process is zero-depleted

$\{Y_n\}$ i.i.d $\sim \text{Exp}(\lambda)$, then $\{N(t)\}$ is Markovian (memoryless property)
in fact it is Poisson.

Poisson process is the only Markovian renewal process

e.g. T_n to be the n th time we replace a lightbulb. ($T_0 = 0$).

$Y_n = T_n - T_{n-1}$ is the lifetime of the n -th bulb.

$\{Y_n\}_{n=2}^\infty$ are iid, $N(t)$ is # of bulbs replaced by time t .

$\{N(t)\}$ is a renewal process.

ELEMENTARY RENEWAL THM:

For a renewal process, if $P(Y_1 < \infty) = 1$, if "mean interarrival time"

$$\mu := E(Y_2) < \infty, \text{ then (a). } \lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\mu} \text{ w.p. 1}$$

$$(b). \lim_{t \rightarrow \infty} \frac{E(N(t))}{t} = \frac{1}{\mu}$$

• Practice: Let Y_1, Y_2, \dots iid $\sim \text{Unif}[0, 10]$

Let $T_0 = 0$, $T_n = Y_1 + Y_2 + \dots + Y_n$, $n \geq 1$. Let $N(t) = \max\{n \geq 0 : T_n < t\}$

(a). Compute $\lim_{t \rightarrow \infty} N(t)/t$.

(b). Approximate $E(\#\{n \geq 1 : 1234 < T_n < 1236\})$

BLACKWELL RENEWAL THM: For a renewal process, s.p.s $P(Y_1 < \infty)$ &

$\mu = E(Y_2) < \infty$, Y_2 not arithmetic i.e. no $\lambda > 0$ s.t. $P(Y_2 = k) = 1$
Then $\forall h > 0$, $\lim_{t \rightarrow \infty} E[N(t+h) - N(t)] = \frac{h}{\mu}$

Renewal Reward Process

Renewal process $P(Y_1 < \infty) = 1$, mean interarrival time $\mu := E(Y_2) < \infty$

S.p.s k th renewal time T_k , receive an reward (or cost) R_k . $\{R_k\}$ iid

Renewal Reward THM:

$$\lim_{t \rightarrow \infty} \frac{R(t)}{t} = \frac{E[R_k]}{\mu} \text{ w.p. 1}$$

[Car purchases]

~~L~~ $\sim \text{Unif}[0, 10]$. buy new car when broke down or after 5 years
 $0 \leq S \leq 10$. 30 - cost, if broke down b4 sell it, cost 5.

• How many cars will you buy each year?

$$\mu = E(Y_2) = E[\min(L, S)]$$

$$\mu = SP(L > S) + E[L | L \leq S] P(L \leq S) = S[(10-S)/10] + \frac{S}{2} \cdot \frac{S}{10} = S - \frac{S^2}{20}$$

By Elementary Renewal Thm part(a). $\lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\mu} = \frac{1}{S - \frac{S^2}{20}}$

- How many cars will you buy between 562 & 566 years?

$$t=562, t+h=566, h=4.$$

By Blackwell Renewal Thm:

$$\frac{h}{M} = \frac{4}{S - S^2/20}$$

- Long-run average car cost per year?

T_k be the time of purchase of k^{th} car.

$T_k = T_k - T_{k-1}$ be k^{th} interarrival time

$R(t)$ total car cost by t .

$$E(R_t) = 30 + 5P(L \leq S) = 30 + 5 \frac{S}{10} = 30 + \frac{S}{2}$$

$$\mu = E(Y_2) = S - S^2/20$$

$$\text{Renewal Reward Thm: } \lim_{t \rightarrow \infty} \frac{R(t)}{t} = \frac{E(R_t)}{\mu} = \frac{30 + S/2}{S - S^2/20}$$

Discrete-time Markov Chains on Continuous State Spaces

state space \mathbb{X} : any non-empty (probably uncountable) set. e.g. \mathbb{R} .

The one-step transition probabilities are ~~$P(x, A)$~~ for each $x \in \mathbb{X}$ and $A \subseteq \mathbb{X}$

discrete-time, general state space, time-homogeneous MC X_0, X_1, X_2, \dots

where $P(X_0 \in A_0, X_1 \in A_1, \dots, X_n \in A_n)$

$$= \int_{x_0 \in A_0} \nu(dx_0) \int_{x_1 \in A_1} P(x_0, dx_1) \cdots \int_{x_{n-1} \in A_{n-1}} P(x_{n-2}, dx_{n-1}) \int_{x_n \in A_n} P(x_{n-1}, dx_n)$$

simplify: $\int_{x_n \in A_n} P(x_{n-1}, dx_n) = P(x_{n-1}, A_n)$

A stationary distribution for MC is a probability measure $\pi(\cdot)$ on \mathbb{X}

$$\text{s.t. } \pi(A) = \int_{\mathbb{X}} \pi(dx) P(x, A) \quad \forall A \subseteq \mathbb{X}$$

A MC on a general state space \mathbb{X} is ϕ -irreducible if \exists a non-zero (σ -finite) measure ψ on \mathbb{X} s.t. if $\psi(A) > 0 \Rightarrow P_x(T_A < \infty) > 0 \quad \forall x \in \mathbb{X}$

$T_A = \inf \{n \geq 0; X_n \in A\}$ 1st hitting time of subset A .

Mid-term review.

EQUAL PERIOD LEMMA: if a chain is irreducible ~~& there are~~ all states must have the same period.

STAT RECURRENCE LEMMA: stationary distn + irreducible \Rightarrow recurrent

- Let $S = \{1, 2, 3\}$ with $\pi_1 = \frac{1}{2}, \pi_2 = \frac{1}{3}, \pi_3 = \frac{1}{6}$. Find irreducible transition prob. $\{p_{ij}\}_{i,j \in S}$ s.t. π is stationary dist.

MCMC \Rightarrow for $i \neq j$, want $p_{ij} = \frac{1}{2} \min(1, \frac{\pi_j}{\pi_i})$

$$p_{21} = p_{32} = \frac{1}{2}, \quad p_{12} = \frac{1}{3}, \quad p_{23} = \frac{1}{4},$$

Then we satisfy $\sum_j p_{ij} = 1 \quad \forall i \in S$

$$\text{so } p_{11} = \frac{2}{3}, \quad p_{22} = \frac{1}{4}, \quad p_{33} = \frac{1}{2}$$

check $\pi_i p_{ij} = \pi_j p_{ji}$. \checkmark

01. Final.

1. $\begin{array}{c} \downarrow \\ 2 \rightarrow 3 \end{array}$

2. $\begin{array}{c} \frac{1}{2} \downarrow \quad \frac{1}{2} \downarrow \\ \overbrace{2 \rightarrow 3}^{\rightarrow} \end{array}$

3. $f_{12} = p_{12} + p_{11} f_{12}$ $\frac{1}{3} = p_{12} + \frac{1}{3} p_{11}$
 $1 = 3p_{12} + p_{11}$

but $p_{12} + p_{11} = 1$
so impossible

4.

STA 447 / 2006 (circle one), Winter 2012, Mid-Term Test

(February 16, 2012. Duration: 60 minutes. Questions: 2. Pages: 3. Total points: 36.)

Notes: You should fully explain your answers! Continue on the back if necessary. NO AIDS ALLOWED. You may use theorems from class. Point values in [square-brackets].

1. Consider a Markov chain with state space $S = \{1, 2\}$, and transition probabilities $p_{11} = 2/3$, $p_{12} = 1/3$, $p_{21} = 1/4$, and $p_{22} = 3/4$.

- (a) [3] Compute $p_{12}^{(2)}$.

$$(P_{ij}) = \begin{pmatrix} \frac{2}{3} & \frac{1}{3} \\ \frac{1}{4} & \frac{3}{4} \end{pmatrix}$$

$$\begin{aligned} P_{12}^{(2)} &= P_{11}P_{12} + P_{12}P_{22} \\ &= \frac{2}{3} \times \frac{1}{3} + \frac{1}{3} \times \frac{3}{4} \\ &= \frac{2}{9} + \frac{1}{4} = \frac{8+9}{36} = \frac{17}{36} \end{aligned}$$

- (b) [2] Determine whether or not this chain is irreducible.

$\forall i, j \in S, f_{ij} > 0$.

in details: $f_{11} = 1 - f_{22} > 0$
 $f_{21} = f_{12} = 1/3 > 0$

so chain is irreducible.

- (c) [4] Let $\pi_1 = 3/7$ and $\pi_2 = 4/7$. Prove that $\{\pi_i\}$ is a stationary probability distribution for this chain.

~~$\pi_i P_{12} = P$~~

~~$\sum_{j=1,2} \pi_j P_{1j} = \pi_1 P_{12} + \pi_2 P_{11} = \frac{3}{7} \times \frac{1}{3} + \frac{4}{7} \times \frac{2}{3}$~~

~~$= \frac{1+2}{7}$~~

reversability

$$\pi_i P_{ij} = \pi_j P_{ji}$$

$$\frac{3}{7} \times \frac{1}{3} = \frac{4}{7} \times \frac{1}{4}$$

$$\pi_1 P_{12} + \pi_2 P_{22} = \pi_2$$

$$\frac{3}{7} \times \frac{1}{3} + \frac{4}{7} \times \frac{3}{4} = \frac{4}{7}$$

OR:

defn:

~~$\pi_i = \sum_j \pi_j P_{ji}$~~

1. (cont'd) Recall that $S = \{1, 2\}$, $p_{11} = 2/3$, $p_{12} = 1/3$, $p_{21} = 1/4$, and $p_{22} = 3/4$.

(d) [3] Determine whether or not $f_{11} = 1$.

$$\begin{aligned} f_{11} &= p_{11} + p_{12}f_{21} = \frac{2}{3} + \frac{1}{3} = 1 \\ f_{21} &= p_{21} + p_{22}f_{21} \\ (1-p_{22})f_{21} &= p_{21} \Rightarrow f_{21} = \frac{p_{21}}{1-p_{22}} = \frac{\frac{1}{4}}{\frac{3}{4}} = \frac{1}{3} \end{aligned} \quad \boxed{ } \quad \checkmark$$

(e) [3] Determine whether or not $f_{21} = 1$.

同理.

(f) [3] Determine whether or not $\sum_{n=1}^{\infty} p_{21}^{(n)} = \infty$.

irreducible + finite state \Rightarrow all recurrent

$$\downarrow$$
$$\sum_{n=1}^{\infty} p_{21}^{(n)} = \infty$$

(g) [3] Determine whether or not this chain is aperiodic.

~~gcd 6~~ b/c. $p_{11} \neq 0, p_{22} \neq 0$. \checkmark aperiodic \checkmark

(h) [2] Determine whether or not $\lim_{n \rightarrow \infty} p_{12}^{(n)} = \pi_2$.

b/c aperiodic.

by MC convergence Thm:

$$\lim_{n \rightarrow \infty} p_{12}^{(n)} = \pi_2$$

2. Consider a Markov chain with state space $S = \{1, 2, 3\}$, and transition probabilities $p_{11} = p_{12} = p_{22} = p_{23} = p_{32} = p_{33} = 1/2$, with $p_{ij} = 0$ otherwise.

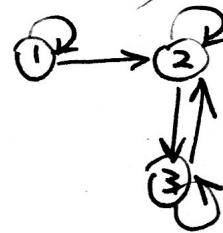
- (a) [3] Determine whether or not this chain is irreducible.

Not irreducible

\nexists $3 \rightarrow 1$.

$$f_{31} = 0$$

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$



- (b) [4] Compute f_{ii} for each $i \in S$.

$$f_{11} = p_{11} + p_{12} f_{21} = \frac{1}{2}$$

$$f_{12} = p_{12} + p_{11} f_{11} \Rightarrow f_{12} = \frac{p_{12}}{1-p_{11}} = 1$$

~~$$f_{13} = p_{13} + p_{12} f_{12}$$~~

$$f_{21} = 0$$

$$f_{22} = p_{22} + p_{23} f_{32} = 1$$

$$f_{23} = p_{23} + p_{22} f_{22} \Rightarrow f_{23} = 1$$

- (c) [3] Specify which states are recurrent, and which states are transient.

similarly: $f_{31} = 0$
 $f_{32} = f_{33} = 1$
 $f_{32} = p_{32} + p_{33} f_{32}$

recurrent: 2, 3

transient: 1

- (d) [3] Compute the value of f_{13} .

$$f_{13} = p_{13} + \cancel{p_{12} f_{23}} + p_{11} f_{13}$$

$$= 0 + \frac{1}{2} \cdot 1 + \frac{1}{2} f_{13}$$

$$\Rightarrow f_{13} = 1$$

[END]

Surname: _____

Given name(s): _____

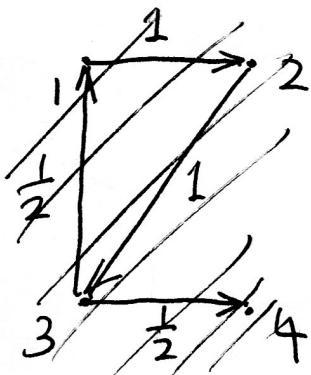
STA 447/2006S, Winter 2008: In-Class Test

(February 28, 2008, 6:10 p.m. Time: 130 minutes.)

(Questions: 6; Pages: 6; Total points: 65.)

NO AIDS ALLOWED – NOT EVEN CALCULATORS.

1. [8 points] Let (p_{ij}) be the transition probabilities for random walk on the graph whose vertices are $V = \{1, 2, 3, 4\}$, with a single edge between each of the four pairs $(1,2)$, $(2,3)$, $(3,1)$, and $(3,4)$, and no other edges. Compute (with full explanation) $\lim_{n \rightarrow \infty} p_{13}^{(n)}$.



$$\pi_{13} = \frac{3}{4 \times 2} = \frac{3}{8} = \lim_{n \rightarrow \infty} P_{13}^{(n)} \text{ by } \text{MCCT}$$

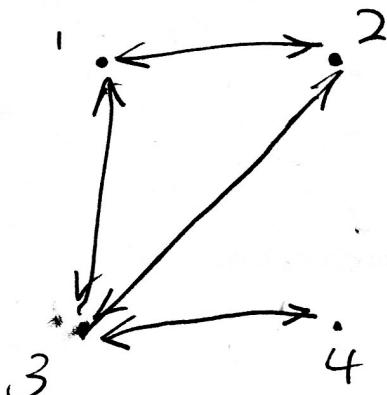
$$\lim_{n \rightarrow \infty} \frac{1}{3} (P_{13}^{(n)} + P_{23}^{(n)} + P_{33}^{(n)}) = \frac{3}{8}$$

irreducible ✓

aperiodic ✓

stationary ? ✓

b/c

 $\pi_1 =$ fixed
b/c reversibility

so by MCCT. ✓

2. Consider the Markov chain with state space $S = \{1, 2, 3\}$, and transition probabilities given by $p_{11} = 1/6$, $p_{12} = 1/3$, $p_{13} = 1/2$, $p_{22} = p_{33} = 1$, and $p_{ij} = 0$ otherwise.

(a) [4 points] Compute (with explanation) f_{12} (i.e., the probability, starting from 1, that the chain will eventually visit 2).

(b) [3 points] Prove that $p_{12}^{(n)} \geq 1/3$, for any positive integer n .

(c) [2 points] Compute $\sum_{n=1}^{\infty} p_{12}^{(n)}$.

(d) [3 points] Relate the answers in parts (a) and (c) to theorems from class about when $f_{ij} = 1$ and when $\sum_{n=1}^{\infty} p_{ij}^{(n)} = \infty$.

3. Let $S = \mathbf{Z}$ (the set of all integers), and let $h : S \rightarrow [0, 1]$ with $\sum_{i \in S} h(i) = 1$. Consider the transition probabilities on S given by $p_{ij} = (1/4) \min(1, h(j)/h(i))$ if $j = i-2, i-1, i+1, i+2$, and $p_{ii} = 1 - p_{i,i-2} - p_{i,i-1} - p_{i,i+1} - p_{i,i+2}$, and $p_{ij} = 0$ whenever $|j - i| \geq 3$.

(a) [10 points] Assuming that $h(i) > 0$ for all i , prove that $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = h(j)$ for all $i, j \in S$. (Carefully justify each step.)

$$P_{ij} = \begin{cases} \frac{1}{4} \min(1, \frac{h(j)}{h(i)}) & \text{if } |i-j| \leq 2 \\ 0 & \text{if } |i-j| \geq 3 \end{cases}$$

treat $h(i)$ as T_i

irreducible: state $i \rightarrow j$ if $0 \leq |i-j| \leq 2$

aperiodic: $\gcd(\overbrace{1, 2, 3, \dots}^{\text{1}}, \underbrace{1, 2, 3, \dots}_{\text{2}}) = 1$. $\gcd(2, 3)$ or $\gcd(1, \dots) = 1$

stationary:

$$h(i)p_{ij} = h(j)p_{ji}$$

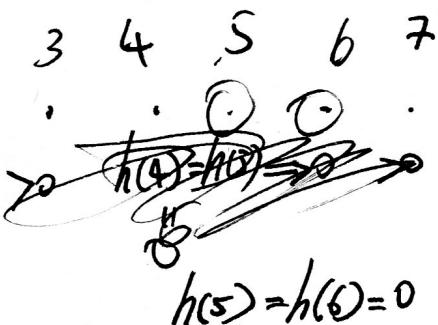
if $|j-i| \geq 3$, trivial

$$\text{if } |j-i| \leq 2, h(i) \cdot \frac{1}{4} \min(1, \frac{h(j)}{h(i)}) = \frac{1}{4} h(j)$$

$$h(j) \cdot \frac{1}{4} \min(1, \frac{h(i)}{h(j)})$$

~~trivial~~ for p_{ii} $h(i) =$

(b) [5 points] Show by example that part (a) might be false if we do not assume that $h(i) > 0$ for all i . [For definiteness, we take $\min(1, h(j)/h(i)) \equiv 1$ whenever $h(i) = 0$.]



~~ex. 3.3~~

$$P_{35} = P_{45} = P_{46} = 0$$

$$p_{ij} = 0 \quad \text{if } i \leq 4 \text{ and } j = 5$$

$$P_{47}^{(n)} = 0 \quad \forall n.$$

but $h(7) > 0$.

4. Consider a Markov chain $\{X_n\}$ with state space $S = \{1, 2, 3, 4, 5\}$, $X_0 = 4$, and transition probabilities specified by $p_{11} = p_{55} = 1$, $p_{21} = 5/7$, $p_{24} = p_{25} = 1/7$, $p_{31} = p_{32} = p_{33} = p_{34} = p_{35} = 1/5$, and $p_{43} = p_{45} = 1/2$. Let $T = \min\{n \geq 1 : X_n = 1 \text{ or } 5\}$.

- (a) [8 points] Determine (with full explanation) whether or not $\{X_n\}$ is a martingale.

$$\sum_j j P_{ij} = i \quad \forall i$$

$$i=1, \quad 1 \times p_{11} + 2 \times p_{12} + 3 \times p_{13} + 4 \times p_{14} + 5 \times p_{15} = 1$$

$$i=2, \quad 1 \times p_{21} + 2 \times p_{22} + 3 \times p_{23} + 4 \times p_{24} + 5 \times p_{25} = \frac{1}{7} \times 4 + \frac{1}{7} \times 5 + \frac{5}{7} = 2$$

$$i=3, \quad \frac{1}{5}(15) = 3$$

$$i=4, \quad 3 \times \frac{1}{2} + 5 \times \frac{1}{2} = 4$$

$$i=5, \quad 5 \quad \checkmark$$

Martingale ✓.

- (b) [4 points] Compute $\mathbf{P}(X_T = 5)$. [Hint: part (a) might help.]

~~$E(X_T = 5) = E(X_0) = 4$~~

$$E(X_T) = 5 \cdot P(X_T = 5) + 1 \cdot P(X_T = 1) = E(X_0) = 4$$

$$P(X_T = 5) = \frac{4 - P(X_T = 1)}{5}$$

$$= \frac{4 - (1 - P(X_T = 5))}{5}$$

$$= \frac{3 + P}{5}$$

$$5P = 3 + P$$

$$4P = 3$$

$$P(X_T = 5) = \frac{3}{4}$$

5. Consider a Markov chain $\{X_n\}$ on the state space $S = \{0, 1, 2, 3, \dots\}$, with $X_0 = 100$, and $p_{ij} = 1/(2i+1)$ if $0 \leq j \leq 2i$, otherwise $p_{ij} = 0$.

(a) [5 points] Prove that $\{X_n\}$ is a martingale. (You may assume without proof that $E|X_n| < \infty$ for all n .)

~~i=0~~ ~~1~~ 0 ✓

~~i=1~~

$$\begin{aligned} & \cancel{P_{11}} + \cancel{P_{12}} + \cancel{P_{13}} \\ &= \frac{1}{2 \times 1 + 1} + \frac{1}{2 \times 2 + 1} \times 2 = \frac{1}{3} + \frac{2}{5} - \frac{11}{15} \\ \sum p_{ij} \cdot j &= \sum_{j=0}^{2i} \frac{j}{2i+1} = \frac{(2i+1)2i/2}{2i+1} = i \quad \checkmark \end{aligned}$$

(b) [5 points] Prove that $\mathbf{P}(\exists n \geq 1 : X_n = 1000) < 1/6$. [Hint: the martingale maximal inequality might help.]

6. Let $\{N(t)\}_{t \geq 0}$ be a Poisson process with rate $\lambda > 0$.

(a) [6 points] Compute the conditional probability $q_\lambda \equiv \mathbf{P}(N(4) = 1 | N(5) = 3)$.

(b) [2 points] Compute $q_{2\lambda} / q_\lambda$. (That is, determine the fraction by which the probability in part (a) changes if we replace λ by 2λ .)

[END]