

Lecture 15

Math 285B

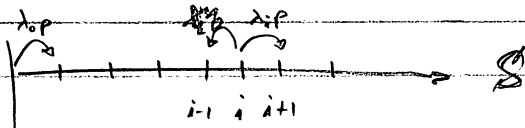
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Class Notes

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Example from Monday



Discrete time skeleton: recurrent iff $\rho \leq q \implies$ CTS time MC does not explode if $\rho \leq q$.

Also from last time: By using detailed balance

$$\pi_i = \pi_0 \left(\frac{\lambda_0}{\lambda_i} \right) \left(\frac{\rho}{q} \right)^i \quad i=0,1,\dots$$

this will be a stationary distribution if

$$\sum_i \frac{1}{\lambda_i} \left(\frac{\rho}{q} \right)^i < \infty.$$

Case 2: Assume, $\lambda_i = z^i$, $i=0,1,2,\dots$

$$\therefore \pi_i = \pi_0 \left(\frac{\rho}{zq} \right)^i \quad i=0,1,2,\dots \quad (\lambda_0=1)$$

We know that $\sum_i \left(\frac{\rho}{zq} \right)^i < \infty$ iff $\frac{\rho}{zq} < 1$

From discrete-time MC, we are guaranteed non-explosion if $\frac{\rho}{q} \leq 1$.

For $\frac{\rho}{q} \leq 1$, we have non-explosion \implies existence of stationary distribution π . By basic limit thm, this implies pos. recurrence and π will be a steady state distribution.

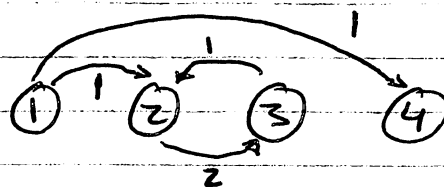
(note: discrete time skeleton is null recurrent for $\rho=q$)

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For $1 < \frac{p}{q} < 2$: discrete time skeleton is transient
 and so its time MC is also transient
 (\therefore not recurrent nor pos. recurrent)
 (In fact, its time MC ^{can} explode when $1 < \frac{p}{q} < 2$).

Another Example:

$$Q = \begin{pmatrix} -2 & 1 & 0 & 1 \\ 0 & -2 & 2 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$



Communicating Classes: $\{1\}, \{2, 3\}, \{4\}$

notes: no arrows leading into 1.

no arrows out of set $\{2, 3\}$

$\{1\}$ - pos. probability of leaving and then once you leave you never come back. \therefore Transient.

$\{2, 3\}$ - no arrows out of $\{2, 3\}$. \therefore Not Transient, hence recurrent (finitely many states \Rightarrow pos. recurrent.)

$\{4\}$ - clearly ~~non~~ recurrent, since starting there means staying there forever.

Can determine stationary distribution for each pos. recurrent communicating class:

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$$\{4\}: \pi_4^{\{4\}} = 1 \quad \left(\pi_i^{\{4\}} = 0 \quad i=1, 2, 3 \right)$$

$$\{2, 3\}: \pi_2^{\{2,3\}}, \pi_3^{\{2,3\}} \text{ non-zero } \left(\pi_i^{\{2,3\}} = 0, \quad i=1, 4 \right)$$

Example (cont'd):

$$\begin{aligned}
 -2\pi_2^{\{2,3\}} + 1\pi_3^{\{2,3\}} &= 0 & \Rightarrow & \pi_3^{\{2,3\}} = 2\pi_2^{\{2,3\}} \\
 2\pi_2^{\{2,3\}} - 1\pi_3^{\{2,3\}} &= 0 & & \\
 & & & \pi_2^{\{2,3\}} + \pi_3^{\{2,3\}} = 1
 \end{aligned}$$

$$\therefore \boxed{\pi_2^{\{2,3\}} = \frac{1}{3} ; \pi_3^{\{2,3\}} = \frac{2}{3}}$$

So, what happens to "poor" state 1?

If we start in 2,3,4 can determine

$$\lim_{t \rightarrow \infty} P_{ij}(t) \text{ for any } i, j$$

If we start in state 1,

$$P_{ij}(t) = P(x(t)=j \mid x(0)=1, J_1 \leq t, x(J_1) \text{ being in same communicating class as } j) \\
 \times P(J_1 \leq t, x(J_1) \text{ is in same cc as } j \mid x(0)=1)$$

where $J_1 = \inf\{t \geq 0 : x(t) \neq 1\}$ and $\underline{j \neq 1}$.

By the strong Markov Property for X:

$$P_{ij}(t) = P(x(t=J_1)=j \mid x(0)=\text{place where } X \text{ jumped to @ time } J_1, \text{ is in same cc. class as } j) \\
 \times P(J_1 \leq t \mid x(0)=1) \cdot P(\text{jump from 1 to cc containing } j \text{ when jump})$$

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Example (cont'd):

So, what happens when we take limit as $t \rightarrow \infty$?
(assume $j=3$).

$$\begin{aligned}\lim_{t \rightarrow \infty} P_{13}(t) &= \lim_{t \rightarrow \infty} P(X(t=J_1)=3 | X(0)=2) (P_{12}) \\ &= \pi_3^{\{2,3\}} = \frac{1}{2}\end{aligned}$$

This concludes cts MC class coverage!

Conditional Expectation:

Let (Ω, \mathcal{F}, P) be a probability space.

$\Omega \rightarrow$ set of possible outcomes

$\mathcal{F} \rightarrow$ σ -algebra (events)

$P \rightarrow$ Probability measure on our space.

Let X be a random variable: $X: \Omega \rightarrow \mathbb{R}$.

(for any interval (a,b) : $-\infty < a < b < \infty$,

$[\omega: X(\omega) \in (a,b)]$ is an event, i.e. an element of \mathcal{F}).

Assume $E[|X|] < \infty$ (here E is expected value under P).

Let \mathcal{G} be a sub- σ -algebra of \mathcal{F} .

(a collection of events in \mathcal{F} that is closed under countable unions & complements).

+ contains \emptyset + Ω .)

Def'n: The conditional expectation of X given \mathcal{G} is a random variable Y such that

(i.) Y is \mathcal{G} -measurable

(i.e. for any $-\infty < a < b < \infty$
 $\{\omega \in \Omega : Y(\omega) \in (a, b)\} \in \mathcal{G}$)

(ii.) for any $A \in \mathcal{G}$

$$E[Y \cdot 1_A] = E[X \cdot 1_A]$$

Luckily, there is such a conditional expectation and it is unique up to a.s. equivalence.

(i.e. if $Y + Y'$ are two such conditional expectations then $P(Y = Y') = 1$.)

Example: Coin Flipping.

Ω = outcomes of two coin flips
= $\{(HH), (TT), (HT), (TH)\}$.

HH	HT
TH	TT

\mathcal{F} = all possible subsets of Ω (\emptyset, Ω are events)

For fair coins, independent:

$$P(HH) = P(TT) = P(HT) = P(TH) = \frac{1}{4}$$

\mathcal{G} corresponds to information associated with seeing the first coin flip. ~~an~~ outcome.

Conveniently we have a partition, $\mathcal{P} = \{\{HH, HT\}, \{TH, TT\}\}$
• \mathcal{G} is generated by taking all unions of sets in \mathcal{P} $\therefore \mathcal{G} = \{\emptyset, \Omega, \{HH, HT\}, \{TH, TT\}\}$.

Example (cont'd):

$X =$ outcome of second coin flip: $\begin{cases} \text{if heads gives } 1 \\ \text{if tails gives } 0 \end{cases}$

Notation: $E[X|G]$ used to denote conditional expected value of X given G .

Compute $Y = E[X|G] \rightarrow$ will be constant on each set in partition \mathcal{P}

Fix $A = \{HH, HT\} \in \mathcal{P}$
Want $E[Y|A] = E[X|A]$ (1)

Let $y_1 =$ value of Y on $\{HH, HT\}$,

$$(1) \Rightarrow y_1 P(A) = E[X|A]$$

$$\Rightarrow y_1 = \frac{E[X|A]}{P(A)} = \frac{1 \cdot \frac{1}{4}}{\frac{1}{2}} = \frac{1}{2}$$

Similarly, $A_2 = \{TH, TT\} \Rightarrow Y = \frac{1}{2}$ on A_2