

Lecture 6. Discrete Time Markov Chain (1)

Sim, Min Kyu, Ph.D.
mksim@seoultech.ac.kr



Markov Chain - Motivation

- I drink a bottle of soda everyday. I drink either Coke or Pepsi everyday. When I choose what to drink for today, I only consider what I drank yesterday. *Markov property*
discrete time *{ State Space }*
- Suppose I drank Coke yesterday, then chance of drinking Coke again today is 0.7.
(What is chance of drinking Pepsi today then?)
- Suppose I drank Pepsi yesterday, then chance of drinking Pepsi again today is 0.5.
(What is chance of drinking Coke today then?)
- In a tabular form?

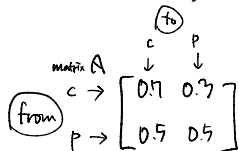
Today \ yesterday	C	P
C	.7	.3
P	.5	.5

DTMC transition diagram.

- How would you describe this situation in diagram?



- How would you represent this situation to mathematical form?



DTMC transition matrix

- 1) 1×1 matrix
- 2) all elements $\in [0, 1]$
- 3) row sum = 1 ... Stochastic matrix

$$S = \{c, p\}$$

$$|S| = \# \text{ elem in } S$$

$$P(X_{n+1} = p | X_n = c) = 0.3 = A_{cp}$$

- Suppose I do this for an year. Which brand of soda I will drink more? (Pepsi or Coke?) *Coke*
- If I drink Coke today, what is the probability that I will be drinking Pepsi on three days later?

0	1	2	3
C	C	C	
	P	P	P

.7 .7 .3
 .7 .3 .5
 .3 .5 .3
 .3 .5 .5

$$P[X_3=P \mid X_0=C]$$

- If I do this for 10 years (3650 days) from now, then how many days I will drink Pepsi?
- Suppose Pepsi is \$1 and Coke is \$1.5. How much on average I spend on soda in a month?
- Suppose that there are a billion customers (who has the same type of consumption pattern) like me in the world. You are working for Pepsi's marketing department and wish to boost Pepsi \rightarrow Pepsi probability from 0.5 to 0.6. On average, how much revenue will be additionally generated by this marketing effort for a day?

Definitions

News vendor

Time 0 : make decision

Time 1 : demand is realized
revenue is determined

1 period : Stochastic Model

Definition - Stochastic Process

■ Stochastic Process: Time + Random

- **Discrete** Time Stochastic Process

- Discrete Time + Random

- $\{X_n : n \geq 0, n \in \mathbb{N}\} = \{X_0, X_1, \dots\}$ index \Rightarrow Time

- **Continuous** Time Stochastic Process

- Continuous Time + Random

- $\{X_t : t \geq 0, t \in \mathbb{R}^+\} = \{X_0, \dots, X_{\infty}, \dots\}$

Definitions - State and State space

- State: value of X_n .

- State space (S): a set of all possible values that X_n can take.

Definitions

Definitions - Markov Property

- The nearest future only depends on the present.
- For a **discrete time** stochastic process $\{X_n : n \geq 0, n \in \mathbb{N}\}$,
 - X_{n+1} depends only on the state of X_n .
 - X_{n+1} is function of X_n and some randomness. $X_{n+1} = f(X_n, \text{random})$
 - $\mathbb{P}(X_{n+1} = j | \underbrace{X_0 = i_0, X_1 = i_1, \dots}_{\text{outdated info irrelevant history}}, \underbrace{X_n = i}_{\text{most recent history}}) = \mathbb{P}(X_{n+1} = j | X_n = i)$
 $= \mathbb{P}(X_{n+1} = j | X_n = i)$
 $= \mathbb{P}(X_{n+1} = j | X_n = i)$

Definition - DTMC

- A **discrete time** stochastic process with **Markov Property** is said to be a DTMC.

Definition - Transition probability (matrix)

- **Transition probability** p_{ij} is the probability that governs “transition” within the state space.

(1 step) • $p_{ij} = \mathbb{P}(X_{n+1} = j | X_n = i) = \mathbb{P}(X_n = j | X_{n-1} = i) = \mathbb{P}(X_1 = j | X_0 = i)$

- **Transition probability matrix** \mathbf{P} is collection of p_{ij}

- $\mathbf{P} = [p_{ij}]$

$$\mathbf{P} \in \mathbb{R}^{|\mathcal{S}| \times |\mathcal{S}|}, \quad |\mathcal{S}| \text{ is \# elem in set } \mathcal{S}$$

Definition - Initial Distribution

- The information of where the chain starts at time 0.
- $a_0 :=$ distribution of X_0 as a row vector.

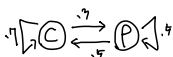
$$\begin{aligned} X_0 = \text{coke} &\Leftrightarrow \mathbb{P}(X_0 = c) = 1 \Leftrightarrow a_0 = (1 \ 0) \\ &\mathbb{P}(X_0 = p) = 0 \\ &\mathbb{P}(X_2 = c) = 0.6 \Leftrightarrow a_2 = (0.6 \ 0.4) \\ &\mathbb{P}(X_2 = p) = 0.4 \end{aligned}$$

TIME INDEX

Formularize Coke & Pepsi MC

■ State Space $\mathcal{S} = \{c, p\}$

■ Transition probability Matrix $P = \begin{bmatrix} .7 & .3 \\ .5 & .5 \end{bmatrix}$

■ Transition Diagram 

■ Initial Distribution

$$a_0 = (\underbrace{.6}_{P(X_0=c)}, \underbrace{.4}_{P(X_0=p)})$$

1-step transition

- Suppose $\mathbb{P}(X_0 = c) = 0.6$ and $\mathbb{P}(X_0 = p) = 0.4$, then what is $\mathbb{P}(X_1 = c)$?

$$X_0 \rightarrow X_1$$

$$\begin{array}{l} c \rightarrow c : \underline{.6} \times \underline{.7} = \underline{.42} \\ p \rightarrow c : \underline{.4} \times \underline{.5} = \underline{.20} \end{array} \quad \} \quad \Sigma = .62 = \mathbb{P}(X_1 = c)$$

$$\mathbb{P}(X_1 = c) = \mathbb{P}(X_1 = c, X_0 = c) + \mathbb{P}(X_1 = c, X_0 = p) \quad \text{cf) } \mathbb{P}(A) = \mathbb{P}(A \cap B) + \mathbb{P}(A \cap B^c)$$

$$= \mathbb{P}(X_1 = c | X_0 = c) \mathbb{P}(X_0 = c) + \mathbb{P}(X_1 = c | X_0 = p) \mathbb{P}(X_0 = p)$$

$$= 0.7 \times \underline{0.6} + 0.5 \times \underline{0.4} = (\underline{0.6} \quad \underline{0.4}) \begin{pmatrix} \underline{0.7} \\ \underline{0.5} \end{pmatrix}$$

$$= 0.62$$

$$a_0 = (.6, .4) \quad P = \begin{bmatrix} .7 & .3 \\ .5 & .5 \end{bmatrix}$$

$$a_0 P = (.6, .4) \begin{pmatrix} .7 & .3 \\ .5 & .5 \end{pmatrix} = (.62, .38) = (\mathbb{P}(X_1 = c), \mathbb{P}(X_1 = p)) = a_1$$

$$\Rightarrow a_1 = a_0 P$$

$$a_2 = a_1 P = (a_0 P) P = a_0 P^2$$

$$a_3 = a_2 P = (a_0 P^2) P = a_0 P^3$$

$$\vdots$$

$$a_n = a_0 P^n = a_1 P^{n-1}$$

Math Review 14 - Probabilities - properties (Revisited)

- $\mathbb{P}(E_1 \cup E_2) = \mathbb{P}(E_1) + \mathbb{P}(E_2)$, „ if $E_1 \cap E_2 = \emptyset$.
- $\mathbb{P}(E) = \mathbb{P}(E \cap F) + \mathbb{P}(E \cap F^c)$ „ $\because (E \cap F) \cap (E \cap F^c) = \emptyset$
- If $F_1 \cap F_2 \cap \dots \cap F_n = \emptyset$ and $\mathbb{P}(F_1 \cup \dots \cup F_n) = 1$, then
$$\mathbb{P}(E) = \sum_{i=1}^n \mathbb{P}(E \cap F_i)$$

$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$, then A^k

$$A = P \Lambda P^{-1}$$

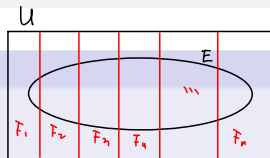
$$\Rightarrow A^k = P \Lambda^k P^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^k \xrightarrow{\infty} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^k \xrightarrow{\infty} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Math Review 15 - Bayes' rule (Revisited)

Definition - Bayes' rule

- Suppose F_1, \dots, F_n are mutually exclusive events
 - $F_i \cap F_j = \emptyset$, or, $\mathbb{P}(F_i \cap F_j) = 0$, for any $i \neq j$
 - $\mathbb{P}(F_1 \cup \dots \cup F_n) = 1$
- In other words, exactly one of the events F_1, \dots, F_n will occur.
- Then, the following holds.

mathematical partition.



$$\begin{aligned}\mathbb{P}(E) &= \mathbb{P}(E \cap F_1) + \dots + \mathbb{P}(E \cap F_n) \\ &= \sum_{i=1}^n \mathbb{P}(E \cap F_i) \\ &= \sum_{i=1}^n \mathbb{P}(E|F_i)\mathbb{P}(F_i)\end{aligned}$$

2-step transition

- Suppose $\mathbb{P}(X_0 = c) = 0.6$ and $\mathbb{P}(X_0 = p) = 0.4$, then what is $\mathbb{P}(X_2 = c)$?

$$\begin{aligned}\mathbb{P}(X_2 = c) &= \mathbb{P}(X_2 = c, X_1 = c) + \mathbb{P}(X_2 = c, X_1 = p) \\ &= \mathbb{P}(X_2 = c | X_1 = c) \mathbb{P}(X_1 = c) + \mathbb{P}(X_2 = c | X_1 = p) \mathbb{P}(X_1 = p) \\ &= 0.7 \times 0.62 + 0.5 \times 0.38\end{aligned}$$

$$a_0 = \begin{pmatrix} 0.6 & 0.4 \end{pmatrix} \quad a_1 p = a_0 p^2 = \begin{pmatrix} \mathbb{P}(X_2 = c) & \mathbb{P}(X_2 = p) \end{pmatrix}$$

$$p = \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix}$$

Transitions in DTMC

$$a_0 \xrightarrow{*P} a_1 \xrightarrow{*P} a_2 \rightarrow \dots \xrightarrow{*P} a_n$$

$$a_n = a_0 P^n$$

$$a_{n+1} = a_n P$$

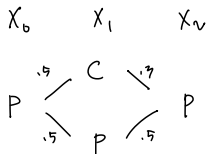
$$\begin{aligned} a_{n+k} &= a_n P^k \\ &= a_k P^n \\ &= a_{n+1} P^{k-1} \end{aligned}$$

2-step transition

- Suppose $X_0 = p$, then what is $\mathbb{P}(X_2 = p)$?
- (Or, equivalently, what is $\mathbb{P}(X_2 = p | X_0 = p)$)?

$$X_0 = p \rightarrow \mathbb{P}(X_1 = c) = 0.5 \quad \mathbb{P}(X_2 = p) = \mathbb{P}(X_2 = p | X_1 = c) \mathbb{P}(X_1 = c) \\ \rightarrow \mathbb{P}(X_1 = p) = 0.5 \quad + \mathbb{P}(X_2 = p | X_1 = p) \mathbb{P}(X_1 = p)$$

$$a_2 = a_0 P^2 = (0 \ 1) \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix}^2 = (0.6 \ 0.4) \underbrace{}_{\mathbb{P}(X_2 = p)}$$



$$.5 \times .7 = .15$$

$$\mathbb{P}(X_2 = p | X_0 = p) \\ = .15 + .25 = .4$$

$$.5 \times .5 = .25$$

2-step transition

- Suppose $X_0 = p$, then what is $\mathbb{P}(X_2 = p)$?
- (Or, equivalently, what is $\mathbb{P}(X_2 = p | X_0 = p)$)?

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"If I only had an hour to chop down a tree, I would spend the first 45 minutes sharpening my axe. - A. Lincoln"