

## *Lecture 6. Discrete Time Markov Chain (1)*

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## Markov Chain - Motivation

discrete time

State Spaces

- 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*
- 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?

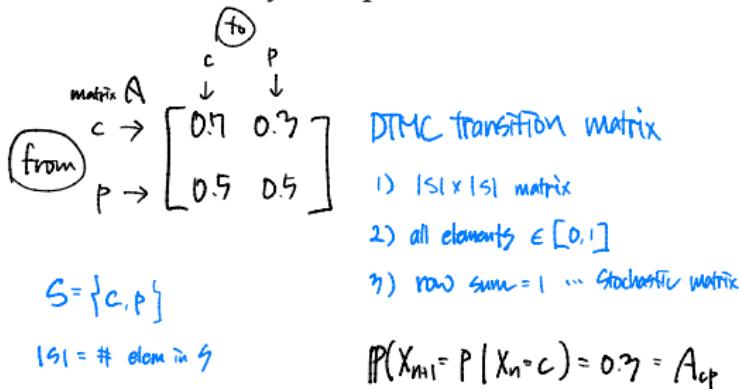
		Today	C	P	
		Yesterday	C	P	
Yesterday	C	.7	.3		
	P	.5	.5		

DTMC transition diagram.

- How would you describe this situation in diagram?



- How would you represent this situation to mathematical form?



- 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	.7 .7 .3
P	P	P	.7 .3 .5 .3 .5 .3 .3 .5 .5

$$\mathbb{P}[X_3 = P \mid X_0 = C]$$

- 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 → 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

NewsVendor

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

- $\{X_t : t \geq 0, t \in \mathbb{R}^+\} = \{X_0, \dots, X_{t_1}, \dots, X_t, \dots\}$

## 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.  $X_{n+1} = f(X_n, \text{random})$
  - $\mathbb{P}(X_{n+1} = j | X_0 = i_0, X_1 = i_1, \dots, X_n = i) = \mathbb{P}(X_{n+1} = j | X_n = i)$   
outdated info  
irrelevant history      most recent  
history  
 $= \mathbb{P}(X_{n+1} = j | X_n = i)$   
 $= \mathbb{P}(X_4 = j | X_3 = i)$

### 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  $p_{ij}$

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

$\mathbf{P} \in \mathbb{R}^{|\mathcal{S}| \times |\mathcal{S}|}$ ,  $|\mathcal{S}|$  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.

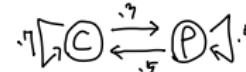
$$X_0 = \text{coca} \Leftrightarrow \mathbb{P}(X_0 = \text{c}) = 1 \Leftrightarrow a_0 = (1 \ 0)$$

$$\mathbb{P}(X_0 = \text{p}) = 0 \quad \text{time index}$$

$$\mathbb{P}(X_0 = \text{c}) = 0.6 \Leftrightarrow a_0 = (0.6 \ 0.4)$$

$$\mathbb{P}(X_0 = \text{p}) = 0.4$$

## Fomularize Coke & Pepsi MC

- State Space  $\mathcal{S} = \{C, P\}$
- Transition probability Matrix  $P = \begin{bmatrix} .7 & .3 \\ .5 & .5 \end{bmatrix}^C_P$
- Transition Diagram 
- Initial Distribution  $\alpha_0 = (\underline{.6}, \underline{.4})$   
 $\begin{array}{l} \text{P}(X_0 = P) \\ \text{P}(X_0 = C) \end{array}$

## 1-step transition

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

$$X_0 \rightarrow X_1$$

$$\begin{aligned} C \rightarrow C &: \underline{.6 \times .7} = .42 \\ P \rightarrow C &: \underline{.4 \times .5} = .20 \end{aligned}$$

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

$$\begin{aligned} &= \mathbb{P}(X_1 = c | X_0 = c) \mathbb{P}(X_0 = c) + \mathbb{P}(X_1 = c | X_0 = p) \mathbb{P}(X_0 = p) \\ &= 0.7 \times \underline{0.6} + 0.5 \times \underline{0.4} = (0.6 \ 0.4) \begin{pmatrix} 0.7 \\ 0.5 \end{pmatrix} \\ &= 0.62 \end{aligned}$$

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

$$a_0 P = (.6 \ 0.4) \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix} = (.62 \ 0.38) = (\mathbb{P}(X_1 = c), \mathbb{P}(X_1 = p)) = a_1$$

$$\Rightarrow a_1 = a_0 P$$

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

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

⋮

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

## Math Review 14 - Probabilities - properties (Revisited)

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

$A = ( \quad )$ , then  $A^k$

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

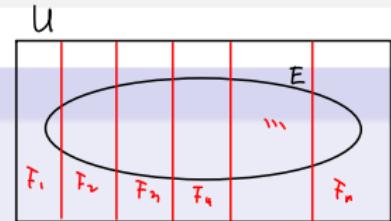
$$\Rightarrow A^k = P \wedge^k P^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^k \xrightarrow{\infty} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}^k \xrightarrow{\infty} \begin{pmatrix} 1 & 0 \\ 0 & \infty \end{pmatrix}$$

## 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  $\mathbb{P}(X_2 = c)$ ?

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

$$a_0 = \begin{pmatrix} 0.6 & 0.4 \\ 0.62 & 0.38 \end{pmatrix} \quad a_1 P = a_0 P^2 = \begin{pmatrix} \mathbb{P}(X_2 = c) & \mathbb{P}(X_2 = p) \end{pmatrix}$$

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

## Transitions in DTMC

$$a_0 \xrightarrow{xP} a_1 \xrightarrow{xP} a_2 \xrightarrow{xP} \dots \xrightarrow{xP} a_n$$

$$a_n = a_0 P^n$$

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

$$a_{n+k} = a_n P^k$$

$$= a_k P^n$$

$$= a_{n+1} P^{k-1}$$

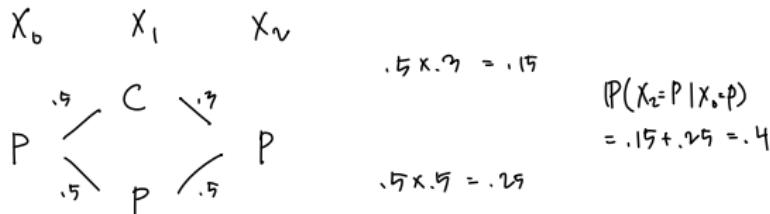
## 2-step transition

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

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

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$$A_2 = A_0 P^2 = (0 \ 1) \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix}^2 = (0.6 \ \underbrace{0.4}_{\mathbb{P}(X_2 = p)})$$



## 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"