

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

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## *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.
- 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?

- 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?)
- If I drink Coke today, what is the probability that I will be drinking Pepsi on three days later?

- 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

### Definition - Stochastic Process

- Stochastic Process: Time + Random
  - Discrete Time Stochastic Process
    - Discrete Time + Random
    - $\{X_n : n \geq 0, n \in \mathbb{N}\}$
  - Continuous Time Stochastic Process
    - Continuous Time + Random
    - $\{X_t : t \geq 0, t \in \mathbb{R}^+\}$

### 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.
  - $\mathbb{P}(X_{n+1} = j | X_0 = i_0, X_1 = i_1, \dots, X_n = i) = \mathbb{P}(X_{n+1} = j | X_n = i)$

### 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.
  - $p_{ij} = \mathbb{P}(X_{n+1} = j | X_n = i) = \mathbb{P}(X_n = j | X_{n-1} = i) = \mathbb{P}(X_1 = j | X_0 = i)$
- **Transition probability matrix**  $\mathbf{P}$  is collection of  $p_{ij}$ 
  - $\mathbf{P} = [p_{ij}]$

### *Definition - Initial Distribution*

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

## Fomularize Coke & Pepsi MC

- State Space
- Transition probability Matrix
- Transition Diagram
- Initial Distribution

## 1-step transition

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

## 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  
$$\mathbb{P}(E) = \sum_{i=1}^n \mathbb{P}(E \cap F_i)$$

## 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.

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

## 2-step transition

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

## *Transitions in DTMC*

## 2-step transition

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

## 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"