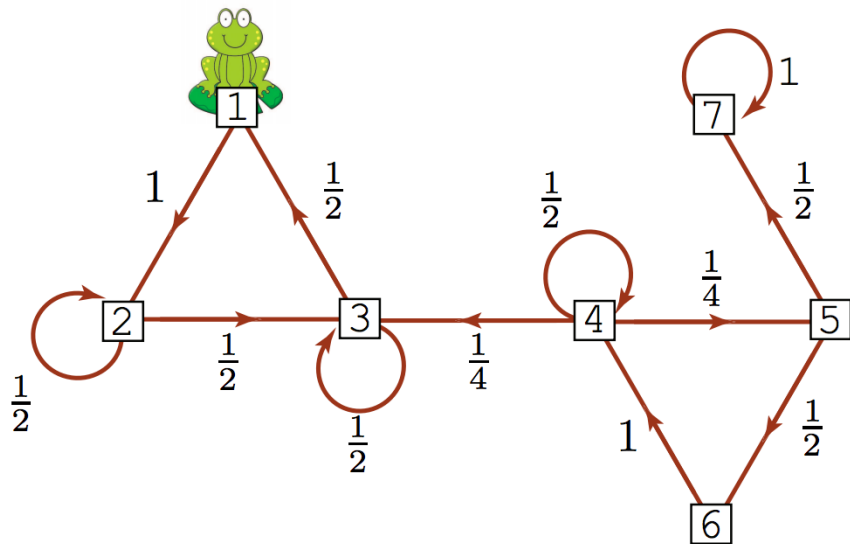


# Markov Chains

# Markov Chains – introduction

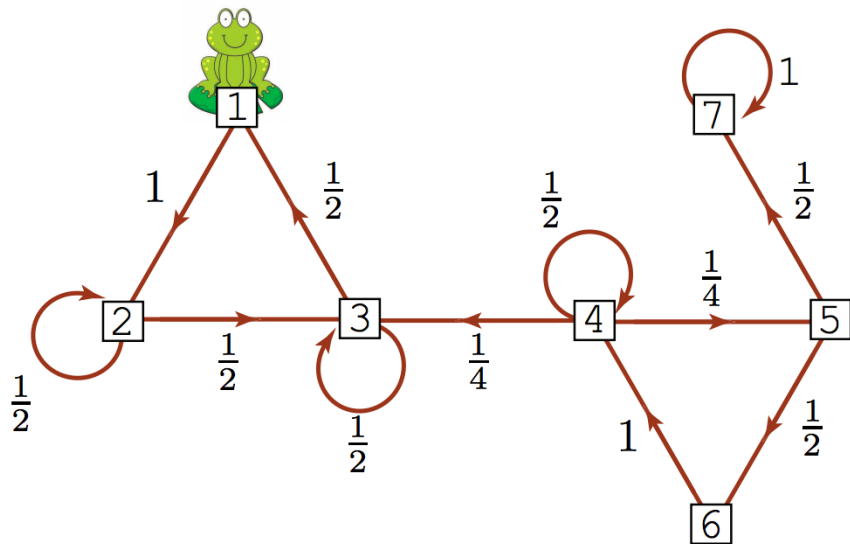
- ❖ Introduced in 1906 by Andrei Andreyevich Markov (1856–1922)
- ❖ (Example) A frog hops about on 7 lily pads. The numbers next to arrows show the probabilities with which, at the next jump, he jumps to a neighboring lily pad (and when out-going probabilities sum to less than 1 he stays where he is with the remaining probability).



Transition matrix

$$P = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{4} & \frac{1}{2} & \frac{1}{4} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

# Markov Chains – introduction



$$P = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{4} & \frac{1}{2} & \frac{1}{4} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

❖ How long does it take to get there?

❖ What happens in the long run?

- Starting in state 1, what is the probability that we are still state 1 after 3 steps? ( $P_{1,1}^{(3)} = \frac{1}{4}$ )
- Starting in state 4, what is the probability that we ever reach state 7?

# Markov Chains – definitions

- ❖ Let  $I = \{i_0, i_1, i_2, \dots, i_{n-1}, i, j\}$  where  $i_k \in I$  ( $i_k$  is called a **state** and  $I$  the **state-space**)
  - ❖ Stochastic process
    - $X_n = i$  : the process is said to be in state  $i$  at time  $n$
  - ❖ Markov Chain
    - Stochastic process where future state  $X_{n+1}$  is independent of the past states and depends only on present state  $X_n$  (discrete state space)
    - $P_{i,j} = P\{X_{n+1} = j | X_n = i, X_n = i_{n-1}, \dots, X_0 = i_0\} = P\{X_{n+1} = j | X_n = i\}$
    - $P_{i,j} \geq 0, \sum_{j=0}^{\infty} P_{i,j} = 1, i = 0, 1, 2, \dots$
- Time homogeneous

## Markov Chains – examples

- ❖ Example 4.1 (Forecasting the weather). Suppose that if it rains today, then it will rain tomorrow with probability  $\alpha$ ; and if it does not rain today, it will rain tomorrow with probability  $\beta$ 
  - Transition probabilities with two states ? (denote “rain” as state 0 and “not rain” 1)
  - $P_{0,0} =$
  - $P_{0,1} =$
  - $P_{1,0} =$
  - $P_{1,1} =$
- ❖ Example 4.2 (A communication system). Two digits 0 and 1. There is a probability  $p$  that the digit entered will be unchanged when it leaves.
  - $P_{0,0} = P_{1,1} =$
  - $P_{0,1} = P_{1,0} =$

## Markov Chains – examples

- ❖ Example 4.3 State space = {Cheerful, So-so, Glum}
  - $(C, C) = 0.5 = P_{0,0}$ ,  $(C, S) = 0.4 = P_{0,1}$ ,  $(C, G) = 0.1 = P_{0,2}$
  - $(S, C) = 0.3 = P_{1,0}$ ,  $(S, S) = 0.4 = P_{1,1}$ ,  $(S, G) = 0.3 = P_{1,2}$
  - $(G, C) = 0.2 = P_{2,0}$ ,  $(G, S) = 0.3 = P_{2,1}$ ,  $(G, G) = 0.5 = P_{2,2}$
  - $P =$

# Markov Chains – examples

## ❖ Example 4.4 (Transforming a process into a Markov Chain)

- O: rain, X: not rain

- 0:  $(O, O) \rightarrow (O, O)$

- 1:  $(X, O) \rightarrow$

- 2:  $(O, X) \rightarrow$

- 3:  $(X, X) \rightarrow$

- $P_{0,0} = P_{1,1} =$

- $P_{0,1} = P_{1,0} =$

# Markov Chains – examples

## ❖ Example 4.5 (A random walk model)

- State space =  $\{0, \pm 1, \pm 2, \pm 3, \dots\}$
- $X_i = \begin{cases} 1 & \text{with probability } p \\ -1 & \text{with } (1 - p) \end{cases}$
- $S_n = X_1 + X_2 + \dots + X_n$
- $P_{i,i+1} =$
- $P_{i,i-1} =$
- $P =$



# Markov Chains – examples

## ❖ Example 4.6 (A gambling model)

- State space =  $\{0, 1, 2, 3, \dots, N\}$
- $X_i = \begin{cases} 1 & \text{with probability } p \\ -1 & \text{with } (1 - p) \end{cases}$
- $P_{0,0} = 1$
- $P_{N,N} = 1$
- $P_{i,i+1} =$
- $P_{i,i-1} =$
- $P =$
- $\{0, N\}$  called absorbing states

# Markov Chains – examples

## ❖ Example 4.7 (Bonus Malus system)

|       |                | Next state if |         |          |                 |
|-------|----------------|---------------|---------|----------|-----------------|
| State | Annual Premium | 0 claims      | 1 claim | 2 claims | $\geq 3$ claims |
| 1     | 200            | 1             | 2       | 3        | 4               |
| 2     | 250            | 1             | 3       | 4        | 4               |
| 3     | 400            | 2             | 4       | 4        | 4               |
| 4     | 600            | 3             | 4       | 4        | 4               |

- $a_k = e^{-\lambda} \frac{\lambda^k}{k!}, k \geq 0$
- $P =$