

# Markov Chains

# Toolbox

- Search: uninformed/heuristic
- Adversarial search
- Probability
- Bayes nets
  - Naive Bayes classifiers
- Statistical inference

# Reasoning over time

- In a Bayes net, each random variable (node) takes on one specific value.
  - Good for modeling static situations.
- What if we need to model a situation that is changing over time?

# Example: Comcast

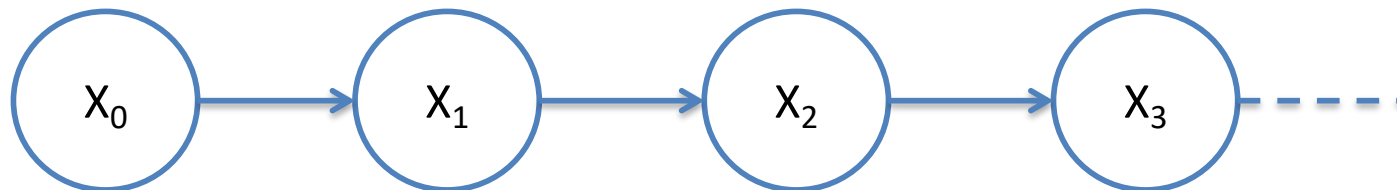
- In 2004 and 2007, Comcast had the worst customer satisfaction rating of any company or gov't agency, including the IRS.
- I have cable internet service from Comcast, and sometimes my router goes down. If the router is online, it will be online the next day with  $\text{prob}=0.8$ . If it's offline, it will be offline the next day with  $\text{prob}=0.4$ .
- How do we model the probability that my router will be online/offline tomorrow? In 2 days?

# Example: Waiting in line

- You go to the Apple Store to buy the latest iPhone. Every minute, the first person in line is served with prob=0.5.
- Every minute, a new person joins the line with probability
  - 1 if the line length=0
  - $\frac{2}{3}$  if the line length=1
  - $\frac{1}{3}$  if the line length=2
  - 0 if the line length=3
- How do we model what the line will look like in 1 minute? In 5 minutes?

# Markov Chains

- A Markov chain is a type of Bayes net with a potentially infinite number of variables (nodes).
- Each variable describes the state of the system at a given point in time ( $t$ ).



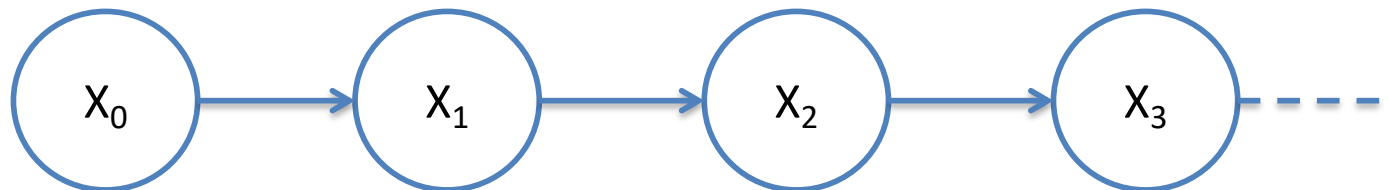
# Markov Chains

- Markov property:

$$P(X_t \mid X_{t-1}, X_{t-2}, X_{t-3}, \dots) = P(X_t \mid X_{t-1})$$

- Probabilities for each variable are identical:

$$P(X_t \mid X_{t-1}) = P(X_1 \mid X_0)$$



# Markov Chains

- Since these are just Bayes nets, we can use standard Bayes net ideas.
  - Shortcut notation:  $X_{i:j}$  will refer to all variables  $X_i$  through  $X_j$ , inclusive.
- Common questions:
  - What is the probability of a specific event happening in the future?
  - What is the probability of a specific sequence of events happening in the future?



# An alternate formulation

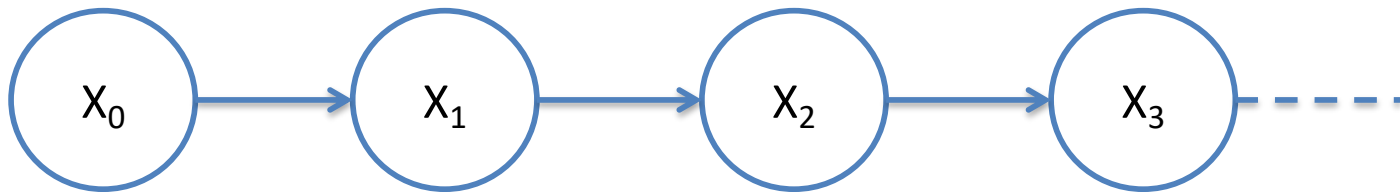
- We have a set of states,  $S$ .
- The Markov chain is always in *exactly one* state at any given time  $t$ .
- The chain transitions to a new state at each time  $t+1$  based only on the current state at time  $t$ .

$$p_{ij} = P(X_{t+1} = j \mid X_t = i)$$

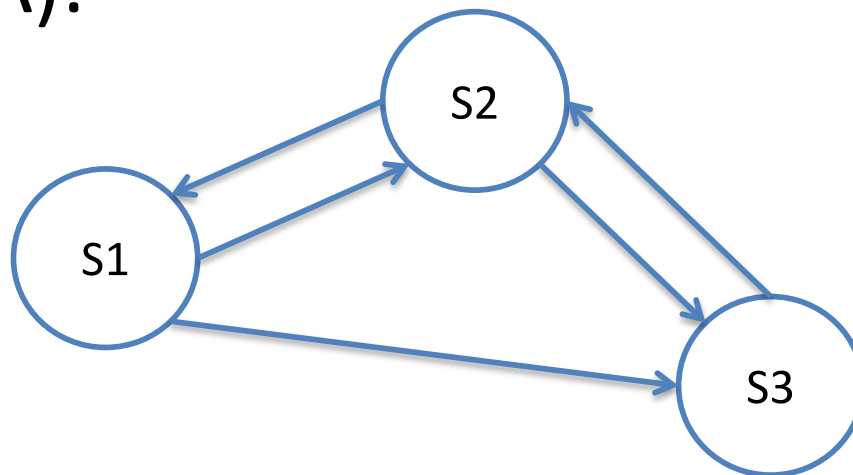
- Chain must specify  $p_{ij}$  for all  $i$  and  $j$ , and starting probabilities for  $P(X_0 = j)$  for all  $j$ .

# Two different representations

- As a Bayes net:



- As a state transition diagram (similar to a DFA/NFA):



# Formulate Comcast in both ways

- I have cable internet service from Comcast, and sometimes my router goes down. If the router is online, it will be online the next day with  $\text{prob}=0.8$ . If it's offline, it will be offline the next day with  $\text{prob}=0.4$ .
- Let's draw this situation in both ways.
- Assume on day 0, probability of router being down is 0.5.

# Comcast

- What is the probability my router is offline for 3 days in a row (days 0, 1, and 2)?
  - $P(X_2=\text{off}, X_1=\text{off}, X_0=\text{off})?$
  - $P(X_2=\text{off} | X_0=\text{off}, X_1=\text{off}) * P(X_0=\text{off}, X_1=\text{off})$      *[mult rule]*
  - $P(X_2=\text{off} | X_0=\text{off}, X_1=\text{off}) * P(X_1=\text{off} | X_0=\text{off}) * P(X_0=\text{off})$
  - $P(X_2=\text{off} | X_1=\text{off}) * P(X_1=\text{off} | X_0=\text{off}) * P(X_0=\text{off})$
  - $p_{\text{off,off}} * p_{\text{off,off}} * P(X_0=\text{off})$

$$P(x_{0:t}) = P(x_0) \prod_{i=1}^t P(x_i | x_{i-1})$$

# More Comcast

- Suppose I don't know if my router is online right now (day 0). What is the prob it is offline tomorrow?

- $P(X_1=\text{off})$

- $P(X_1=\text{off}) = P(X_1=\text{off}, X_0=\text{on}) + P(X_1=\text{off}, X_0=\text{off})$

- $P(X_1=\text{off}) = P(X_1=\text{off} | X_0=\text{on}) * P(X_0=\text{on})$   
+  $P(X_1=\text{off} | X_0=\text{off}) * P(X_0=\text{off})$

$$P(X_{t+1}) = \sum_{x_t} P(X_{t+1} | x_t) P(x_t)$$

# More Comcast

- Suppose I don't know if my router is online right now (day 0). What is the prob it is offline **the day after tomorrow?**

- $P(X_2=\text{off})$

- $P(X_2=\text{off}) = P(X_2=\text{off}, X_1=\text{on}) + P(X_2=\text{off}, X_1=\text{off})$

- $P(X_2=\text{off}) = P(X_2=\text{off} | X_1=\text{on}) * P(X_1=\text{on})$   
+  $P(X_2=\text{off} | X_1=\text{off}) * P(X_1=\text{off})$

$$P(X_{t+1}) = \sum_{x_t} P(X_{t+1} | x_t) P(x_t)$$

# Markov chains with matrices

- Define a transition matrix for the chain:

$$T = \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix}$$

- Each row of the matrix represents the transition probabilities **leaving** a state.
- Let  $v_t$  = a row vector representing the probability that the chain is in each state at time  $t$ .
- $v_t = v_{t-1} * T$

# Formulate this matrix

- If the stock market is up one day, then it will be up the next day with  $\text{prob}=0.7$ .
- If it's down one day, it will be down the next day with  $\text{prob}=0.4$ .



# Mini-forward algorithm

- Suppose we are given the value of  $X_t$  or a probability distribution over  $X_t$  and we want to predict  $X_{t+1}, X_{t+2}, X_{t+3}\dots$
- Make row vector  $v_t = P(X_t)$ 
  - Note that  $v_t$  can be something like  $[1, 0]$  if you know the true value of  $X_t$ , or it can be a distribution over values.
- $v_{t+1} = v_t * T$
- $v_{t+2} = v_{t+1} * T = v_t * T * T = v_t * T^2$
- $v_{t+3} = v_t * T^3$
- $v_{t+n} = v_t * T^n$

# Back to the Apple Store...

- You go to the Apple Store to buy the latest iPhone.
- Every minute, a new person joins the line with probability
  - 1 if the line length=0
  - $\frac{2}{3}$  if the line length=1
  - $\frac{1}{3}$  if the line length=2
  - 0 if the line length=3
- Immediately after (in the same minute), the first person is helped with prob = 0.5
- Model this as a Markov chain, assuming the line starts empty. Draw the state transition diagram.
- What is  $T$ ? What is  $v_0$ ?