

Machine Learning Overview Part 2

Lyle Ungar

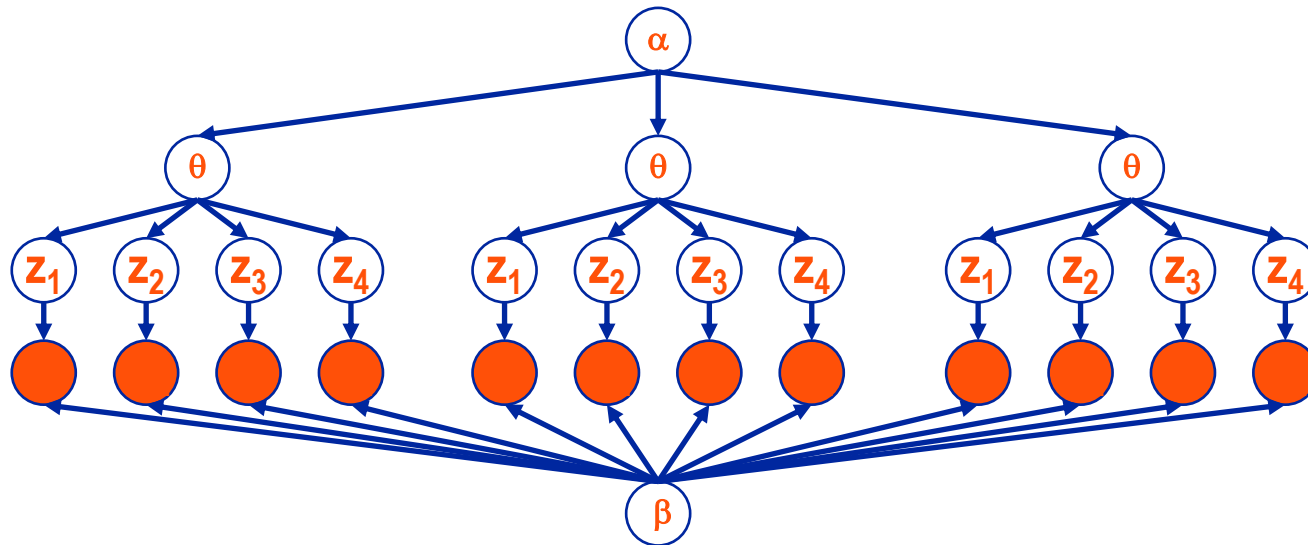
EM

- ◆ When is EM used?
- ◆ Expectation step does what?
- ◆ Maximization step does what?
- ◆ EM is a kind of gradient descent in what?

EM

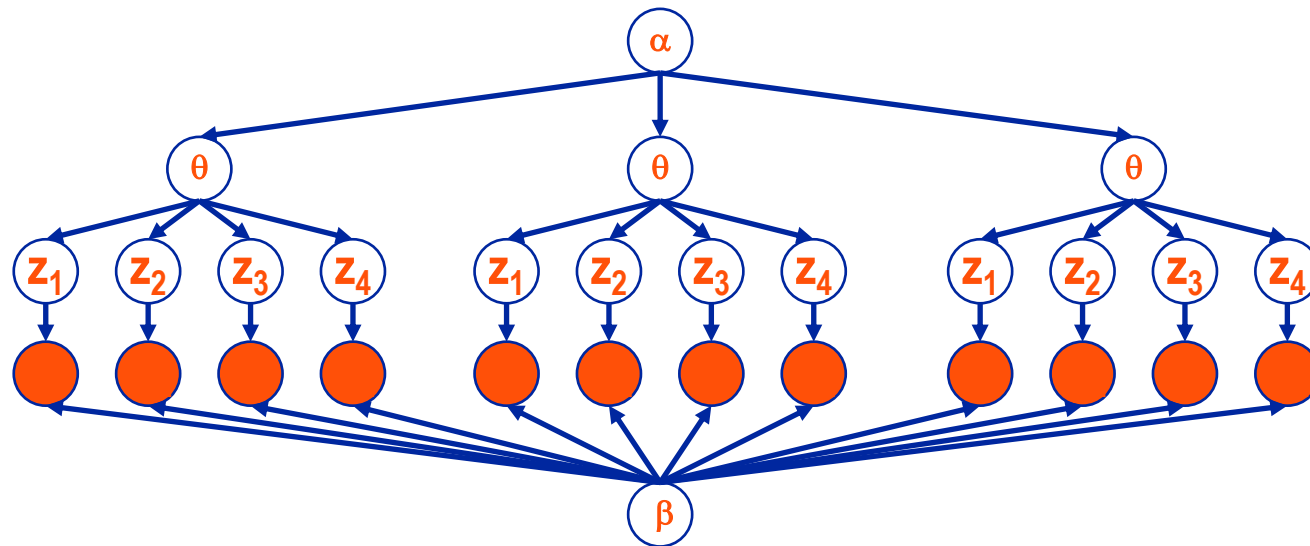
- ◆ **When is EM used?**
 - Unobservable latent variables (GMM, HMM, LDA)
 - Missing data
- ◆ **Expectation step does what?**
 - $E(\mathbf{z})$ – expected value of the missing data
- ◆ **Maximization step does what?**
 - Finds the MLE or MAP of the model parameters
- ◆ **EM is a kind of gradient descent in what?**
 - likelihood

The LDA Model



- ◆ For each document,
 - Choose the topic distribution $\theta \sim \text{Dirichlet}(\alpha)$
 - For each of the N words w_n :
 - Choose a topic $z \sim \text{Multinomial}(\theta)$
 - Then choose a word $w_n \sim \text{Multinomial}(\beta_z)$
 - ◆ Where each topic has a different parameter vector β for the words

The LDA Model



- ◆ Generative or discriminative?
- ◆ What is observed?
- ◆ What is hidden?
- ◆ What are the parameters?
- ◆ What does the E step estimate?
- ◆ What does the M step estimate?

RL: model-based vs. model free

◆ Model-based

- MDP: $p(s_{t+1}|s_t, a_t)$

◆ Model-free

- Q-learning

Model-based RL

◆ Policy iteration

- Policy evaluation: Bellman's equation

$$v_{k+1}(s) = \sum_a \pi(a|s) \sum_{s',r} p(s', r|s, a) [r + \gamma v_k(s')]$$

- Policy improvement

$$\pi'(s) = \operatorname{argmax}_a \sum_{s',r} p(s', r|s, a) [r + \gamma v_\pi(s')]$$

Model-free RL

◆ Q-learning

Choose A from S using policy derived from Q (e.g., ϵ -greedy)

Take action A , observe R, S'

$$Q(S, A) \leftarrow Q(S, A) + \alpha [R + \gamma \max_a Q(S', a) - Q(S, A)]$$

◆ What are

- $Q(s,a)$
- α, γ
- ϵ -greedy?

◆ Is the above on-policy or off-policy?

Update Current V or Q

- ◆ **Look only one step ahead**
 - Q learning = TD(0)
- ◆ **Do randomized search**
 - Monte Carlo
 - Much faster than trying to do exhaustive search
 - alphaGo approach

Auto-encoders

- ◆ What is the most widely used auto-encoder?
- ◆ What loss function do auto-encoders minimize?
- ◆ What prevents an auto-encoder from learning the identity mapping?
- ◆ When do you use a linear vs. a nonlinear auto-encoder?

Auto-encoders

- ◆ **What is the most widely used auto-encoder?**
 - PCA
- ◆ **What loss function do auto-encoders minimize?**
 - Reconstruction error (always L2 in this course)
- ◆ **What prevents an auto-encoder from learning the identity mapping?**
 - Limited number of components, noise, sparsity penalties ...
- ◆ **When do you use a linear vs. a nonlinear auto-encoder?**

Belief nets

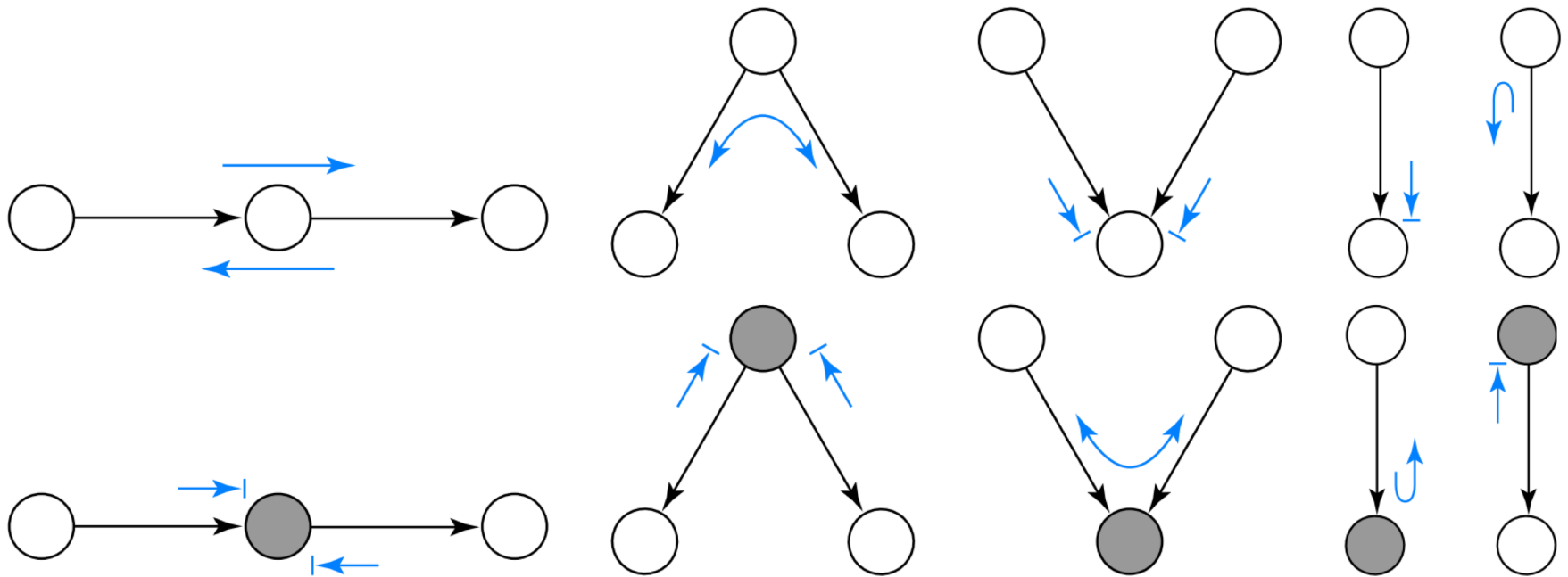
◆ D-separation

- X and Y are d-separated if there is no active path between them.

◆ Active path

Ten rules of Bayes Ball

An undirected path is active if a Bayes ball travelling along it never encounters the “stop” symbol: $\longrightarrow|$



If there are no active paths from X to Y when $\{Z_1, \dots, Z_k\}$ are shaded, then