

# Variational Bayes EM & Methods Review



# EM: reminder

$$\log p(X|\theta) \geq \mathcal{L}(\theta, q) = \mathbb{E}_{q(T)} \log \frac{p(X, T|\theta)}{q(T)} dT \rightarrow \max$$

Marginal likelihood

Variational  
lower bound



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E-step

$$\mathcal{L}(\theta, q) \rightarrow \max_q$$

M-step

$$\mathcal{L}(\theta, q) \rightarrow \max_{\theta}$$



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$$\log p(X|\theta) \geq \mathcal{L}(\theta, q) = \mathbb{E}_{q(T)} \log \frac{p(X, T|\theta)}{q(T)} dT \rightarrow \max$$

Marginal likelihood

Variational  
lower bound

E-step

$$\mathcal{L}(\theta, q) \rightarrow \max_q \Leftrightarrow \mathcal{KL}[q(T) \parallel p(T|X, \theta)] \rightarrow \min_q$$

M-step

$$\mathcal{L}(\theta, q) \rightarrow \max_{\theta} \Leftrightarrow \mathbb{E}_{q(T)} \log p(X, T|\theta) \rightarrow \max_{\theta}$$



# E-step

$$\mathcal{KL}[q(T) \parallel p(T|X, \theta)] \rightarrow \min_q$$

Full posterior

$$q(T) = p(T|X, \theta)$$

Variational inference

$$\mathcal{KL}[q(T) \parallel p(T|X, \theta)] \rightarrow \min_{q \in Q}$$



# Model

**Known:**  $X$  data

**Unknown:**  $\theta$  parameters

**Unknown:**  $T$  latent variables



# Methods

Accurate

- Full inference  $p(T, \theta | X)$

Slow

Inaccurate

Fast



# Methods

Accurate

- Full inference  $p(T, \theta | X)$

- Mean field  $q(T)q(\theta) \approx p(T, \theta | X)$

Inaccurate

Slow

Fast





# Methods

Accurate

- Full inference  $p(T, \theta | X)$
- Mean field  $q(T)q(\theta) \approx p(T, \theta | X)$
- EM algorithm  $q(T), \theta = \theta_{\text{MP}}$

Inaccurate

Slow

Fast



# Methods

Accurate

- Full inference  $p(T, \theta | X)$
- Mean field  $q(T)q(\theta) \approx p(T, \theta | X)$
- EM algorithm  $q(T), \theta = \theta_{\text{MP}}$
- Variational EM  $q_1(T_1) \dots q_d(T_d), \theta = \theta_{\text{MP}}$

Inaccurate

Slow

Fast



# Methods

