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$$= \mathcal{N}(\widetilde{\mu}_i, \widetilde{\Sigma}_i)$$

$$\max_{\theta} \mathbb{E}_{q(T)} \sum_{i} \log p(x_i \mid t_i, \theta) p(t_i)$$

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$$= \sum_{i} \mathbb{E}_{q(t_i)} \log \left( \frac{1}{Z} \exp(\dots) \exp(\dots) \right)$$

$$\max_{\theta} \mathbb{E}_{q(T)} \sum_{i} \log p(x_i \mid t_i, \theta) p(t_i)$$

$$= \sum_{i} \log \frac{1}{Z}$$

$$+ \sum_{i} \mathbb{E}_{q(t_i)} \log (\exp (\dots) \exp (\dots))$$

$$\max_{\theta} \mathbb{E}_{q(T)} \sum_{i} \log p(x_i \mid t_i, \theta) p(t_i)$$

$$= \sum_{i} \log \frac{1}{Z}$$

$$+ \sum_{i} \mathbb{E}_{q(t_i)} \log \left( \exp \left( \dots \right) \exp \left( -\frac{t_i^2}{2} \right) \right)$$

$$\max_{\theta} \mathbb{E}_{q(T)} \sum_{i} \log p(x_i \mid t_i, \theta) p(t_i)$$

$$= \sum_{i} \log \frac{1}{Z}$$

$$+ \sum_{i} \mathbb{E}_{q(t_i)} \log \left( \exp \left( -\frac{(x - Wt_i - b)^2}{2\sigma^2} \right) \exp \left( -\frac{t_i^2}{2} \right) \right)$$

$$\max_{\theta} \mathbb{E}_{q(T)} \sum_{i} \log p(x_i \mid t_i, \theta) p(t_i)$$

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$$at_i^2 + ct_i + d$$

### **Summary**

#### Probabilistic formulation of PCA

- Allows for missing values
- Straightforward iterative scheme for large dimensionalities
- Can do mixture of PPCA
- Hyperparameter tuning (number of components or choose between diagonal and full covariance)