

Proximity Variational Inference

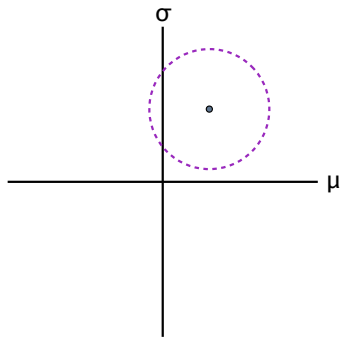
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$$\mathcal{L}(\lambda) = \mathbb{E}_{q(\mathbf{z}; \lambda)}[\log p(\mathbf{x}, \mathbf{z}) - \log q(\mathbf{z}; \lambda)]$$

Gradient ascent using proximity operators

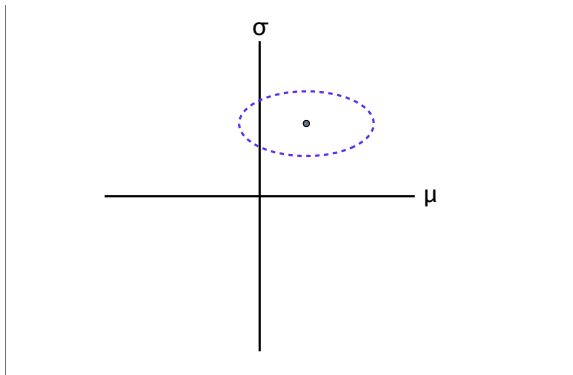
$$U(\boldsymbol{\lambda}_{t+1}) = \mathcal{L}(\boldsymbol{\lambda}_t) + \nabla \mathcal{L}(\boldsymbol{\lambda}_t)^\top (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t) - \frac{1}{2\rho} (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t)^\top (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t)$$

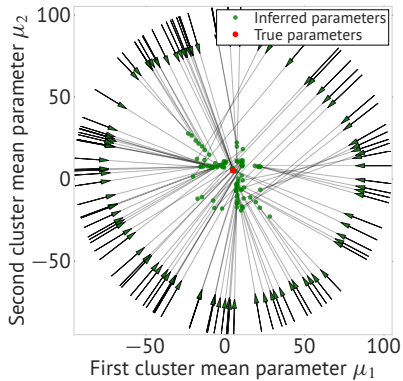
$$\Rightarrow \boldsymbol{\lambda}_{t+1}^* = \boldsymbol{\lambda}_t + \rho \nabla \mathcal{L}(\boldsymbol{\lambda}_t)$$



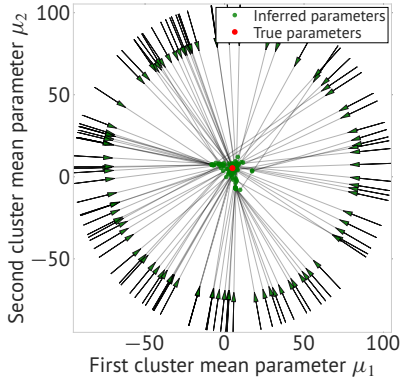
Proximity operators for variational inference

$$\begin{aligned}U(\boldsymbol{\lambda}_{t+1}) &= \mathcal{L}(\boldsymbol{\lambda}_t) + \nabla \mathcal{L}(\boldsymbol{\lambda}_t)^\top (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t) \\ &\quad - \frac{1}{2\rho} (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t)^\top (\boldsymbol{\lambda}_{t+1} - \boldsymbol{\lambda}_t) \\ &\quad - kd(f(\boldsymbol{\lambda}_t), f(\boldsymbol{\lambda}_{t+1}))\end{aligned}$$





Variational Inference



Proximity Variational Inference

Binarized MNIST



(a) Proximity Variational Inference (b) Data (c) Variational Inference