

We derive a lower bound for a linear dynamical system (graphical model is identical to an HMM). Except we want to do variational inference on it. Let \vec{y} represent observations and \vec{x} represent latent variables.

$$\log p(\vec{y}) \geq \int_{\vec{x}} q(\vec{x}|\vec{y}) \log \frac{p(\vec{y}, \vec{x})}{q(\vec{x}|\vec{y})} d\vec{x} \quad (1)$$

where $\vec{x} \sim q(\vec{x}|\vec{y})$, $\vec{z} \sim q(\vec{z}|\vec{x})$. Given that the graphical model induces structural independencies, we can extract the following factorizations:

$$p(\vec{y}|\vec{x}) = \prod_{t=1}^T p(y_t|x_t) \quad (2)$$

$$p(\vec{x}) = p(x_1) \prod_{t=2}^T p(x_t|x_{t-1}) \quad (3)$$

$$q(\vec{x}|\vec{y}) = q(x_1|\vec{y}) \prod_{t=2}^T q(x_t|x_{t-1}, \vec{y}) \quad (4)$$

$$(5)$$

where T is the sequence length. We can write the evidence lower bound as:

$$\mathcal{L} = \mathbb{E}_{q_\phi(\vec{x}|\vec{y})} \left[\log \frac{(\prod_{t=1}^T p_\theta(y_t|x_t))(p_\theta(x_1) \prod_{t=2}^T p_\theta(x_t|x_{t-1}))}{q_\phi(x_1|y_{1:T}) \prod_{t=2}^T q_\phi(x_t|x_{t-1}, y_{1:T})} \right] \quad (6)$$

$$= \mathbb{E}_{q_\phi(\vec{x}|\vec{y})} \left[\log p_\theta(z_1) + \sum_{t=1}^T \log p_\theta(y_t|x_t) + \sum_{t=2}^T \log p_\theta(x_t|x_{t-1}) - \log q_\phi(x_1|y_{1:T}) - \sum_{t=2}^T \log q_\phi(x_t|x_{t-1}, y_{1:T}) \right] \quad (7)$$