

$$P_t(f) \approx \sum_z P(f|z) \cdot P_t(z)$$

The probability of one sound quanta : $P_t(f)$

Suppose sound quanta are ~~iid.~~ independent to each other, we have $\sum_{f,t} V_{ft}$ sound quanta

Data likelihood, i.e. the probability of all sound quanta :

$$\prod_t \prod_f P_t(f)^{V_{ft}}, \quad \text{log-likelihood} \quad \sum_t \sum_f V_{ft} \log P_t(f)$$

ML: Maximize the data likelihood by adjusting the model parameters, i.e. $P(f|z)$ and $P_t(z)$.

For each sound quanta at time t , it has two random variables, f and z . To completely describe a sound quanta, we should use $P_t(f, z)$.

EM algorithm.

$$\text{Complete data likelihood} \quad \prod_t \prod_f P_t(f, z)^{V_{ft}}, \quad \sum_t \sum_f V_{ft} \log P_t(f, z)$$

M step: maximize the expected complete data likelihood.

$$\begin{aligned} L &= E_{P_t(z|f)} \left[\sum_t \sum_f V_{ft} \log P_t(f, z) \right] \\ &= \sum_t \sum_f V_{ft} E_{P_t(z|f)} [\log P_t(f, z)] \\ &= \sum_t \sum_f V_{ft} E_{P_t(z|f)} \left[\log P(f|z) + \log P_t(z) \right] \\ &= \sum_t \sum_f V_{ft} E_{P_t(z|f)} [\log P(f|z)] + \sum_t \sum_f V_{ft} E_{P_t(z|f)} [\log P_t(z)] \\ &= \sum_t \sum_f V_{ft} \sum_z P_t(z|f) \log P(f|z) + \sum_t \sum_f V_{ft} \sum_z P_t(z|f) \log P_t(z) \end{aligned}$$

$$\text{st. } \sum_f P(f|z) = 1 \quad \text{for all } z$$

$$\sum_z P_t(z) = 1 \quad \text{for all } t$$

We can enforce these constraints by using Lagrange multipliers.

$$Q = L + \sum_z \rho_z \left(1 - \sum_f P(f|z) \right) + \sum_t \tau_t \left(1 - \sum_z P_t(z) \right)$$

$$\max_{P(f|z), P_t(z)} Q$$

Take derivatives:

$$\text{for a particular } f \text{ and } z : \frac{\partial Q}{\partial P(f|z)} = \sum_t V_{ft} P_t(z|f) \frac{1}{P(f|z)} + \rho_z = 0$$

$$\Rightarrow \sum_t V_{ft} P_t(z|f) = P(f|z) \rho_z$$

$$\Rightarrow \sum_f \sum_t V_{ft} P_t(z|f) = \sum_f P(f|z) \rho_z = \rho_z$$

$$\Rightarrow P(f|z) = \frac{\sum_t V_{ft} P_t(z|f)}{\rho_z} = \frac{\sum_t V_{ft} P_t(z|f)}{\sum_f \sum_t V_{ft} P_t(z|f)}$$

$$\text{for a particular } z \text{ and } t : \frac{\partial Q}{\partial P_t(z)} = \sum_f V_{ft} P_t(z|f) \frac{1}{P_t(z)} - \tau_t = 0$$

$$\Rightarrow \sum_f V_{ft} P_t(z|f) = P_t(z) \tau_t$$

$$\Rightarrow \sum_z \sum_f V_{ft} P_t(z|f) = \sum_z P_t(z) \tau_t = \tau_t$$

$$\Rightarrow P_t(z) = \frac{\sum_f V_{ft} P_t(z|f)}{\tau_t} = \frac{\sum_f V_{ft} P_t(z|f)}{\sum_z \sum_f V_{ft} P_t(z|f)}$$

$$E \text{ step: } P_t(z|f) = \frac{P_t(z, f)}{P_t(f)} = \frac{P(f|z) P_t(z)}{\sum_z P(f|z) P_t(z)}$$

Compare with NMF - KL multiplicative update:

$$W_{ia} \leftarrow W_{ia} \frac{\sum_\mu H_{a\mu} V_{i\mu} / (WH)_{i\mu}}{\sum_\mu H_{a\mu}}$$

$$H_{a\mu} \leftarrow H_{a\mu} \frac{\sum_i W_{ia} V_{i\mu} / (WH)_{i\mu}}{\sum_k W_{ka}}$$

$$W_{ia} \sim P(f|z) \quad i \sim f, \quad a \sim z.$$

$$H_{a\mu} \sim P_t(z) \quad a \sim z, \quad \mu \sim t.$$

~~E-step~~