

6 - meeting

$\langle \exp(\Phi_{\text{fict}}) \rangle$ is constant.
how to obtain $f(\Phi_{\text{fict}})$ (use hmt probably)

• $\underline{P(\Phi_{\text{fict}})}$ Normal → will contribute $-\frac{1}{2} \Phi_{\text{fict}}^2$

• $P(X_{\text{ft}} | \dots)$ Normal

→ only considering Φ_{fict} (param. of interest)

• Laplace Mean should be mode
Variance should be $-H^{-1}$ ↗ from BP-NMF paper

• $f(\Phi_{\text{fict}}) \propto a \exp(b\Phi) - \frac{1}{2} \Phi^2$

- mean - not closed probability
- variance - pairwise closed

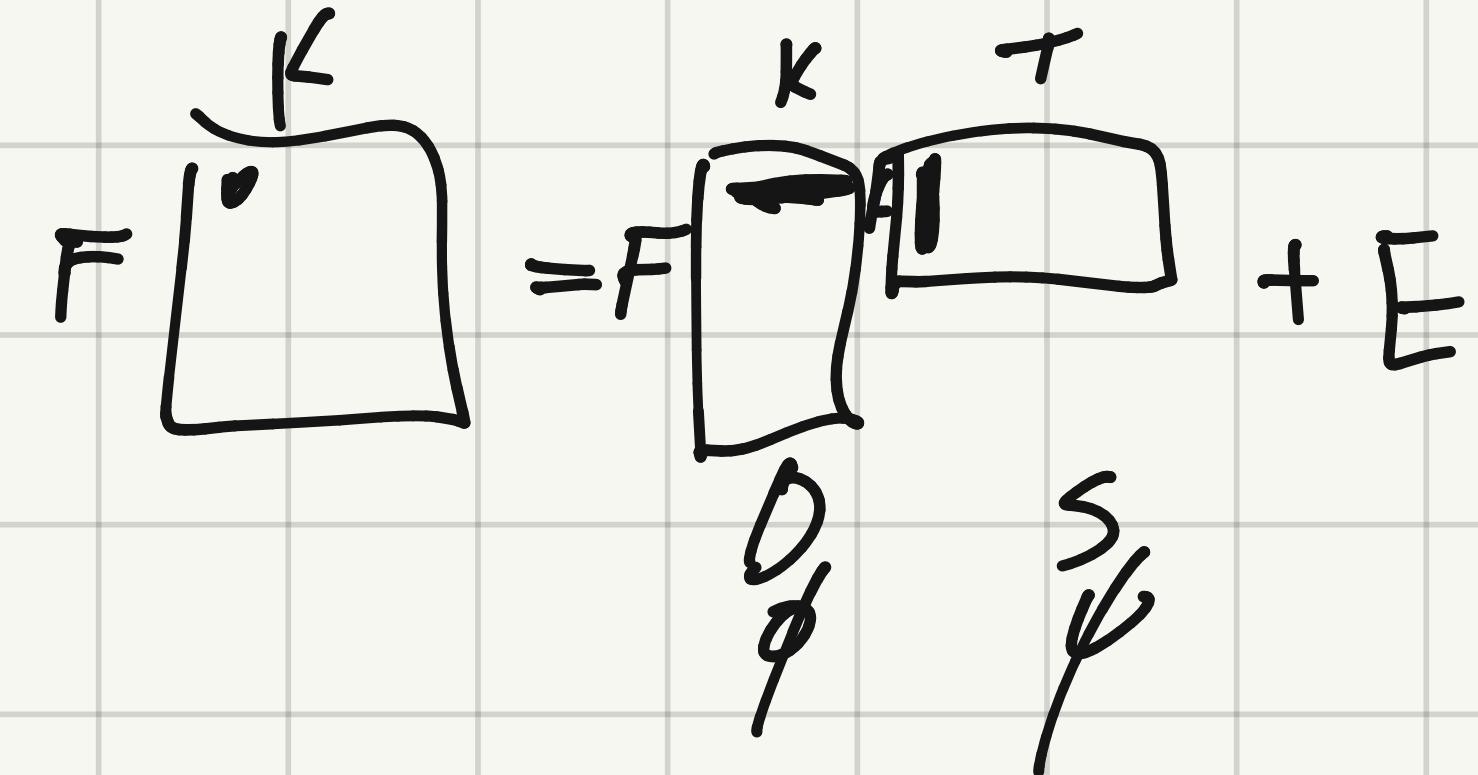
Another paper (referenced in BP-NMF paper)

Bayesian non parametric matrix factorization for recorded music.

• $\theta_l \rightarrow$ shrink to 0

1. Problem Setting

$$\underset{F \times T}{l} = \underset{F \times K}{D} (\underset{K \times T}{S \odot Z}) + \underset{F \times T}{E}$$



$$x_t = D(S_t \odot z_t) + \varepsilon_t$$

$$\log d_k \sim N(0, I_F) \rightarrow \phi$$

$$s_t \sim \Gamma(\alpha, \beta) \rightarrow \psi$$

$$z_k \sim \text{Bern}(\pi_k)$$

$$\pi_k \sim \text{Beta}\left(\frac{a_0}{K}, \frac{b_0(K-1)}{K}\right)$$

$$\varepsilon_t \sim N(0, \tau_{\varepsilon}^{-1} I_F)$$

$$\gamma_{\varepsilon} \sim \Gamma(c_0, d_0)$$

2. Laplace Approximation Variational Inference

$$\Theta = \{D, S, Z, \pi, \gamma_{\varepsilon}\} \rightarrow \{\phi, \psi, z, \pi, \gamma_{\varepsilon}\}$$

$$\phi_{fk} = \log(D_{fk})$$

$$\psi_{kt} = \log(S_{kt})$$

Variational distribution:

$$q(\Theta) = q(\gamma_{\varepsilon}) \prod_{k=1}^K q(\pi_k) \left(\prod_{f=1}^F q(\phi_{fk}) \right) \prod_{t=1}^T \left[q(\psi_{kt}) q(z_{kt}) \right]$$

where

$$q(\phi_{fk}) = N(\mu_{fk}^{(\phi)}, \frac{1}{\gamma_{fk}^{(\phi)}})$$

$$q(\psi_{kt}) = N(\mu_{kt}^{(\psi)}, \frac{1}{\gamma_{kt}^{(\psi)}})$$

$$q(z_{kt}) = \text{Bern}(p_{kt}^{(z)})$$

$$q(\pi_k) = \text{Beta}(\alpha_k^{(\pi)}, \beta_k^{(\pi)})$$

$$q(\gamma_{\varepsilon}) = \Gamma(\alpha^{(\varepsilon)}, \beta^{(\varepsilon)})$$

$$\cdot \text{ELBO} = \mathbb{E}_q [\log P(X, \Theta)] - \mathbb{E}_q [\log q(\Theta)]$$

$$\log P(X) \geq \text{ELBO}$$

$$= \mathbb{E}_q [\log P(X | \phi, \psi, z, \pi, \gamma_{\varepsilon})] + \mathbb{E}_q [\log P(\phi)]$$

$$+ \mathbb{E}_q [\log P(\psi)] + \mathbb{E}_q [\log P(z | \pi)] + \mathbb{E}_q [\log P(\pi)]$$

$$+ \mathbb{E}_q [\log P(\gamma_{\varepsilon})] + H_q[q]$$

entropy

2.1 Update ϕ

$$\begin{aligned}
 \cdot q(\phi_{fk}) &\propto \exp \left\{ \mathbb{E}_{-\phi_{fk}} [\log P(X, \Theta)] \right\} \\
 &\propto \exp \left\{ \mathbb{E}_{-\phi_{fk}} [\log P(x_f | \phi_f, \psi, z, \gamma_\varepsilon)] \right. \\
 &\quad \left. + \mathbb{E}_{-\phi_{fk}} [\log P(\phi_{fk})] \right\} \\
 &= \exp \left\{ \langle \log P(x_f | \phi_f, \psi, z, \gamma_\varepsilon) \rangle_{-\phi_{fk}} + \log P(\phi_{fk}) \right\} \\
 &= \exp \{ f(\phi_{fk}) \}
 \end{aligned}$$

normal(0, 1)

\rightarrow need to compute to apply Laplace approx.

$$\langle \log P(x_f | \phi_f, \psi, z, \gamma_\varepsilon) \rangle = \sum_{t=1}^T \langle \log P(x_{ft} | \phi_f, \psi_t, z_t, \gamma_\varepsilon) \rangle$$

where $P(x_{ft} | \phi_f, \psi_t, z_t, \gamma_\varepsilon)$ can be written as exp. family (gaussian).

$$\begin{aligned}
 \cdot X_{ft} &\sim N \left(\sum_{j=1}^K [D_{fj} (S_{jt} \times Z_{jt})], \frac{1}{\gamma_\varepsilon} \right) \\
 &= N \left(\sum_{j=1}^K [\exp(\phi_{fj}) \exp(\psi_{jt}) \cdot Z_{jt}], \frac{1}{\gamma_\varepsilon} \right)
 \end{aligned}$$

refer to (basis & likelihood from AMSI \rightarrow Exp. family in Zeros), we know the general normal exp. family form. Substituting, we get

$$\begin{aligned}
 \cdot \eta &= \begin{bmatrix} \frac{\mu}{\sigma^2} \\ -\frac{1}{2\sigma^2} \end{bmatrix} = \begin{bmatrix} \gamma_\varepsilon \cdot \sum_{j=1}^K \exp(\phi_{fj}) \cdot \exp(\psi_{jt}) \cdot Z_{jt} \\ -\frac{1}{2} \gamma_\varepsilon \end{bmatrix}
 \end{aligned}$$

$$\begin{aligned}
 \cdot A(\eta) &= \frac{\mu^2}{2\sigma^2} + \log \sigma = -\frac{\eta_1^2}{4\eta_2} - \frac{1}{2} \log(-2\eta_2) \\
 &= -\frac{1}{2} \left[\gamma_\varepsilon \left(\sum_{j=1}^K \exp(\phi_{fj}) \exp(\psi_{jt}) \cdot Z_{jt} \right)^2 + \log(\gamma_\varepsilon) \right]
 \end{aligned}$$

$$\cdot T(X_{ft}) = \begin{bmatrix} x_{ft} \\ x_{ft}^2 \end{bmatrix}$$

(var.)
 ϕ_{fk}

$$\begin{aligned}
 \text{now, } \langle \log P(x_{ft} | \phi_f, \psi_t, z_t, \gamma_\varepsilon) \rangle_{-\phi_{fk}} &\propto \langle \eta^T \rangle T(x_{ft}) - \langle A(\eta) \rangle \\
 &= \langle \gamma_\varepsilon \rangle \left[\sum_{j=1}^K \langle \exp(\phi_{fj}) \exp(\psi_{jt}) \cdot Z_{jt} \rangle \right] x_{ft} - \frac{1}{2} \langle \gamma_\varepsilon \rangle x_{ft}^2 \\
 &\quad - \frac{1}{2} \left[\langle \gamma_\varepsilon \rangle \left\langle \left(\sum_{j=1}^K \exp(\phi_{fj}) \exp(\psi_{jt}) Z_{jt} \right)^2 \right\rangle + \langle \log(\gamma_\varepsilon) \rangle \right]
 \end{aligned}$$

$$= \langle \gamma_e \rangle \left[\sum_{j=1}^K \langle \exp(\phi_{fj}) \rangle \langle \exp(\psi_{jt}) \rangle \langle z_{jt} \rangle \right] X_{ft} - \frac{1}{2} \langle \gamma_e \rangle \times \cancel{\frac{2}{\gamma_e}} \xrightarrow{\text{(no } \phi_{fk} \text{)}} (no \phi_{fk})$$

$$- \frac{1}{2} \left[\langle \gamma_e \rangle \left\langle \left(\sum_{j=1}^K \exp(\phi_{fj}) \cdot \exp(\psi_{jt}) \cdot z_{jt} \right)^2 \right\rangle + \langle \log(\gamma_e) \rangle \right]$$

α

$$\propto \langle \gamma_e \rangle X_{ft} \left[\exp(\phi_{fk}) \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle + \sum_{j \neq k} \cancel{\langle \exp(\phi_{fj}) \rangle \langle \exp(\psi_{jt}) \rangle} \langle z_{jt} \rangle \right] \xrightarrow{\text{(no } \phi_{fk} \text{)}}$$

$$- \frac{1}{2} \langle \gamma_e \rangle [\dots]$$

- $\bullet \left(\sum_{j=1}^K \exp(\phi_{fj}) \exp(\psi_{jt}) z_{jt} \right)^2$

$$= \left(\exp(\phi_{fk}) \exp(\psi_{kt}) z_{kt} \right) + \underbrace{\sum_{j \neq k} \exp(\phi_{fj}) \exp(\psi_{jt}) z_{jt}}_P(X_{ft})^2 P(x_t)$$

α

$$\propto \exp(2\phi_{fk}) \exp(2\psi_{kt}) z_{kt}^2 + 2 \exp(\phi_{fk}) \exp(\psi_{kt}) z_{kt} \times \sum_{j \neq k}$$

$$\Rightarrow \langle (\sum \dots)^2 \rangle$$

$$\propto \exp(2\phi_{fk}) \langle \exp(2\psi_{kt}) \rangle \langle z_{kt}^2 \rangle$$

$$+ 2 \exp(\phi_{fk}) \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \left\langle \sum_{j \neq k} \exp(\phi_{fj}) \exp(\psi_{jt}) z_{jt} \right\rangle$$

now, put $\sum_{t=1}^T$ for $\log P(X_F | \dots)$

$\propto \langle \gamma_e \rangle \exp(\phi_{fk}) \sum_{t=1}^T X_{ft} \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle$

$\quad \quad \quad z_i \sim \text{Bern}(p_i)$

$\quad \quad \quad \Rightarrow z_i^2 \sim \text{Bern}(p_i)$

$\quad - \frac{1}{2} \langle \gamma_e \rangle \left[\exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt}^2 \rangle = \langle z_{kt} \rangle \right]$

$\quad \quad \quad + 2 \exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \left\langle \sum_{j \neq k} \exp(\phi_{fj}) \exp(\psi_{jt}) z_{jt} \right\rangle$

$$= \langle \gamma_e \rangle \exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \left\langle X_{ft} - \sum_{j \neq k} \exp(\phi_{fj}) \exp(\psi_{jt}) z_{jt} \right\rangle$$

$$- \frac{1}{2} \langle \gamma_e \rangle \cdot \exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle$$

$\Rightarrow f(\phi_{fk}) \propto \langle \gamma_e \rangle \exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \langle \hat{x}_{ft}^{-k} \rangle$

$$- \frac{1}{2} \langle \gamma_e \rangle \cdot \exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle - \frac{1}{2} \phi_{fk}^2$$

from $\log P(\phi_{fk})$
Normal

$$\langle r_{\varepsilon} \rangle \sum_{j=1}^K \langle \exp(\phi_{fj}) \times \exp(\psi_{jt}) \times z_{jt} \rangle X_{ft} - \frac{1}{2} \langle r_{\varepsilon} \rangle \times \overset{\text{(no } \phi_{fk})}{X_{ft}^2}$$

$$- \frac{1}{2} \langle r_{\varepsilon} \rangle \left[\left(\sum_{j=1}^K \exp(\phi_{fj}) \cdot \exp(\psi_{jt}) \cdot z_{jt} \right)^2 + \langle (z_{jt})^2 \rangle \right]$$

ϕ_{fk}

$$\propto \langle r_{\varepsilon} \rangle X_{ft} \left[\exp(\phi_{fk}) \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle + \sum_{j \neq k} \langle \exp(\phi_{fj}) \times \exp(\psi_{jt}) \rangle \langle z_{jt} \rangle \right]$$

$$- \frac{1}{2} \langle r_{\varepsilon} \rangle \left[(\exp(\phi_{fk}) \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle)^2 + 2 \exp(\phi_{fk}) \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle \times \sum_{j \neq k} \dots \right]$$

part $\sum_{t=1}^T$ for $\log P(x_f)$

$$\propto \langle r_{\varepsilon} \rangle \exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle X_{ft}$$

\leftarrow different

$$- \frac{1}{2} \langle r_{\varepsilon} \rangle \exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle^2 + 2 \exp(\phi_{fk}) \times \dots$$

$$\Rightarrow f(\phi_{fk}) \propto \langle r_{\varepsilon} \rangle \exp(\phi_{fk}) \times \left[\sum_{t=1}^T \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle \hat{X}_{ft} \right]$$

\leftarrow correct derivation

$$- \frac{1}{2} \langle r_{\varepsilon} \rangle \exp(2\phi_{fk}) \times \left[\sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \right]$$

\leftarrow what should I use
the mean.

$$\text{now, } \frac{\partial f}{\partial \phi_{fk}} = \langle r_{\varepsilon} \rangle \left[\exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle \langle \hat{X}_{ft} \rangle \right]$$

$$- \exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \]$$

$$- \phi_{fk} \leftarrow$$

Newton
Raphson

set 0 to find $\overset{\wedge}{\phi}_{fk}$ (no closed form)

$$\frac{\partial^2 f}{\partial \phi_{fk}^2} = \langle r_{\varepsilon} \rangle \left[\exp(\phi_{fk}) \sum_{t=1}^T \langle \exp(\psi_{kt}) \rangle \langle z_{kt} \rangle \langle \hat{X}_{ft} \rangle \right]$$

$$- 2 \exp(2\phi_{fk}) \sum_{t=1}^T \langle \exp(2\psi_{kt}) \rangle \langle z_{kt} \rangle \right] - 1$$

$\therefore q(\phi_{fk}) \approx N(\mu_{fk}^{(\phi)}, \sigma_{fk}^{(\phi)})$ with (using Laplace approx.)

$$\mu_{fk}^{(\phi)} = \overset{\wedge}{\phi}_{fk}, \quad \sigma_{fk}^{(\phi)} = - \frac{\partial^2 f(\overset{\wedge}{\phi}_{fk})}{\partial \phi_{fk}^2} \text{ update.}$$

2.2. Update ψ

$$\begin{aligned} \cdot q(\psi_{kt}) &\propto \exp\left\{\langle \log P(X, \theta) \rangle_{-\psi_{kt}}\right\} \\ &\propto \exp\left\{\langle \log P(x_t | \phi, \psi_t, z_t, r_\xi) \rangle + \langle \log P(\psi_{kt}) \rangle\right\} \\ &= \exp\left\{f(\psi_{kt})\right\} \end{aligned}$$

here, S_{kt} is Gamma, so ψ_{kt} is log-gamma distributed.

Now, Gamma has pdf: $\frac{\beta^\alpha}{\Gamma(\alpha)} S_{kt}^{\alpha-1} \cdot \exp(-\beta S_{kt})$ but $S_{kt} = \exp(\psi_{kt})$

$$\Rightarrow p(\psi_{kt}) = \frac{\beta^\alpha}{\Gamma(\alpha)} \exp(\psi_{kt})^{\alpha-1} \cdot \exp(-\beta \exp(\psi_{kt}))$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \cdot \exp\left\{\alpha \psi_{kt} - \beta \exp(\psi_{kt})\right\}$$

$$\Rightarrow \langle \log P(\psi_{kt}) \rangle \propto \alpha \psi_{kt} - \beta \exp(\psi_{kt})$$

and $\langle \log P(x_t | \phi, \psi_t, z_t, r_\xi) \rangle$ (same form as before, swapping appropriately)

$$\propto \langle r_\xi \rangle \exp(\psi_{kt}) \sum_{f=1}^F \langle \exp(\phi_{fk}) \rangle \langle z_{kt} \rangle \langle \hat{x}_{ft}^{-K_f} \rangle$$

$$- \frac{1}{2} \langle r_\xi \rangle \exp(2\psi_{kt}) \sum_{f=1}^F \langle \exp(2\phi_{fk}) \rangle \langle z_{kt} \rangle$$

$$\Rightarrow f(\psi_{kt}) \propto \langle r_\xi \rangle \exp(\psi_{kt}) \sum_{f=1}^F \langle \exp(\phi_{fk}) \rangle \langle z_{kt} \rangle \langle \hat{x}_{ft}^{-K_f} \rangle$$

$$- \frac{1}{2} \langle r_\xi \rangle \exp(2\psi_{kt}) \sum_{f=1}^F \langle \exp(2\phi_{fk}) \rangle \langle z_{kt} \rangle$$

$$+ \alpha \psi_{kt} - \beta \exp(\psi_{kt})$$

⋮

Similar update as before, do $\frac{\partial f}{\partial \psi_{kt}}$, $\frac{\partial^2 f}{\partial \psi_{kt}^2}$

2.3 Update z

- $q(z_{kt}) \propto \exp \left\{ \langle \log P(X, \theta) \rangle_{z_{kt}} \right\}$
 $\propto \exp \left\{ \langle \log P(x_t | \phi, \psi_t, z_t, \gamma_\varepsilon) \rangle + \langle \log P(z_{kt} | \pi_k) \rangle \right\}$

Since z_{kt} is Bernoulli, can explicitly compute $P_1 \propto q(z_{kt}=1)$ and $P_0 \propto q(z_{kt}=0)$

- if $z_{kt} = 1$,

from $\langle \log P(x_t | \phi, \psi_t, z_t, \gamma_\varepsilon) \rangle$ (wrt. z_{kt})

$$\propto \langle \gamma_\varepsilon \rangle \sum_{f_{k1}}^F \langle \exp(\psi_{kt}) \rangle \langle \exp(\phi_{fk}) \rangle \langle z_{kt} \rangle \langle \hat{x}_{ft}^{-K_f} \rangle$$

$$- \frac{1}{2} \langle \gamma_\varepsilon \rangle \sum_{f_{k1}}^F \langle \exp(2\psi_{kt}) \rangle \langle \exp(2\phi_{fk}) \rangle \langle z_{kt} \rangle$$

$$\propto \langle \gamma_\varepsilon \rangle \sum_{f_{k1}}^F \{\langle \exp(\phi_{fk}) \rangle \dots\} \quad \text{and} \quad \langle \log P(z_{kt} | \pi_k) \rangle \propto \langle \log \pi_k \rangle$$

- if $z_{kt} = 0$, $\log q(z_{kt}) \propto P_0 = \langle \log(1 - \pi_k) \rangle$

where $\langle \log \pi_k \rangle = \psi(\alpha_k^{(\pi)}) - \psi(\alpha_k^{(\pi)} + \beta_k^{(\pi)})$
 $(\psi(\cdot) \text{ is digamma function})$ (* fact for $E[\log(\text{Beta})]$)

$$\text{so } \langle \log(1 - \pi_k) \rangle = \psi(\beta_k^{(\pi)}) - \psi(\alpha_k^{(\pi)} + \beta_k^{(\pi)})$$

$$\Rightarrow p_{kt}^{(z)} \leftarrow \frac{\exp(p_1)}{\sum_{i \in \Omega} \exp(p_i)} \quad \text{update rule}$$

$$= \exp(p_0) + \exp(p_1)$$

2.4 π_k , 2.5 γ_ε have closed form due to conjugacy
with \uparrow
Beta-Bernoulli normal-gamma