

# Cosine basis

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## 1 Model specifications

$$\begin{aligned}y_i &= w_i^\top \beta + f(x_i) + \epsilon_i, & \epsilon_i &\sim \mathcal{N}(0, \sigma^2) \\ \theta_j | \sigma, \tau, \gamma &\sim \mathcal{N}(0, \sigma^2 \tau^2 \exp[-j\gamma]) \\ \tau^2 &\sim \text{IG}\left(\frac{r_{0,\tau}}{2}, \frac{s_{0,\tau}}{2}\right) \\ \sigma^2 &\sim \text{IG}\left(\frac{r_{0,\sigma}}{2}, \frac{s_{0,\sigma}}{2}\right) \\ \beta &\sim \mathcal{N}(\mu_\beta^0, \Sigma_\beta^0) \\ \gamma &\sim \text{Exp}(\omega_0) \\ |\psi| &= \gamma, \quad \psi \sim \text{DE}(0, \omega_0) \\ \varphi_j(x) &= \sqrt{2} \cos(\pi j x)\end{aligned}$$

Joint density:

$$\begin{aligned}p(y, \Theta) &= \mathcal{N}(y | W\beta + f_J, \sigma^2 I_n) \left\{ \prod_{j=1}^J \mathcal{N}(\theta_j | 0, \sigma^2 \tau^2 \exp[-j|\psi|]) \right\} \text{IG}\left(\tau^2 \middle| \frac{r_{0,\tau}}{2}, \frac{s_{0,\tau}}{2}\right) \text{IG}\left(\sigma^2 \middle| \frac{r_{0,\sigma}}{2}, \frac{s_{0,\sigma}}{2}\right) \mathcal{N}(\beta | \mu_\beta^0, \Sigma_\beta^0) \\ &\quad \text{DE}(\psi | 0, \omega_0)\end{aligned}$$

We will use the joint density to derive the LB and updating algorithm. The variational distributions are

$$\begin{aligned}q_1(\beta) &= \mathcal{N}(\mu_\beta^q, \Sigma_\beta^q) \\ q_2(\theta_J) &= \mathcal{N}(\mu_\theta^q, \Sigma_\theta^q) \\ q_3(\sigma^2) &= \text{IG}\left(\frac{r_{q,\sigma}}{2}, \frac{s_{q,\sigma}}{2}\right) \\ q_4(\tau^2) &= \text{IG}\left(\frac{r_{q,\tau}}{2}, \frac{s_{q,\tau}}{2}\right) \\ q_5(\psi) &= \mathcal{N}(\mu_\psi^q, \sigma_\psi^{2q}) \quad (\text{NCVMP}).\end{aligned}$$

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## 2 Lower bound

### 2.1 LB: $\mathbb{E} [\ln p(y|\Theta)]$

$$\begin{aligned}\mathbb{E} [\ln p(y|\Theta)] &= -\frac{n}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \mathbb{E} \left[ \left( y - W\beta - \varphi_J^\top \theta \right)^\top \left( y - W\beta - \varphi_J^\top \theta \right) \right] \\ &= -\frac{n}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \left( y - W\mu_\beta^q - \varphi_J^\top \mu_\theta^q \right)^\top \left( y - W\mu_\beta^q - \varphi_J^\top \mu_\theta^q \right) - \frac{1}{2} \left( \text{Tr} \left( W^\top W \Sigma_\beta^q \right) + \text{Tr} \left( \varphi_J \varphi_J^\top \Sigma_\theta^q \right) \right)\end{aligned}$$

### 2.2 LB: $\mathbb{E} [\ln p(\theta_j|\sigma, \tau, \psi)]$

$$\sum_{j=1}^J \mathbb{E} [\ln p(\theta_j|\sigma, \tau, \psi)] = \sum_{j=1}^J \mathbb{E} \left[ -\frac{1}{2} \ln(2\pi) + \ln \frac{1}{\sigma} + \ln \frac{1}{\tau} + \frac{j}{2} |\psi| - \frac{\theta_j^2 e^{j|\psi|}}{2\sigma^2 \tau^2} \right]$$

Let's note the following fact: if  $X \sim \mathcal{N}(\mu, \sigma^2)$ , then  $|X| \sim \text{folded-Normal}(\mu, \sigma^2)$ . Then,

$$\begin{aligned}\mathbb{E} |X| &= \sigma \sqrt{\frac{2}{\pi}} \exp \left\{ -\frac{\mu^2}{2\sigma^2} \right\} + \mu \left( 1 - 2\Phi \left( \frac{-\mu}{\sigma} \right) \right) \\ &= \sigma \sqrt{\frac{2}{\pi}} \exp \left\{ -\frac{\mu^2}{2\sigma^2} \right\} - \mu \text{erf} \left( \frac{-\mu}{\sqrt{2}\sigma} \right) \\ \mathbb{E} e^{t|X|} &= \exp \left\{ \frac{\sigma^2 t^2}{2} + \mu t \right\} \left[ 1 - \Phi \left( -\frac{\mu}{\sigma} - \sigma t \right) \right] + \exp \left\{ \frac{\sigma^2 t^2}{2} - \mu t \right\} \left[ 1 - \Phi \left( \frac{\mu}{\sigma} - \sigma t \right) \right].\end{aligned}$$