# July 04, 2019

## 1 Variable Selection

Start with spline model

$$Y = b(X)\beta + \epsilon$$

Select variable? select basis?

- $Y = b(\Gamma X)\beta + \epsilon$
- $Y = b(X)\Gamma\beta + \epsilon$

### 2 Basis Selection Model

$$Y = b(X)\Gamma\beta + \epsilon$$

Let 
$$Z = b(X)$$

#### 2.1 prior

$$\begin{split} Y|\beta,\sigma^2,\Gamma &\sim N(Z\Gamma\beta,\sigma^2\cdot I) \\ \sigma^2 &\sim Inverse - Gamma(a,b) \\ \beta_j|\sigma_{\beta_j}^2 &\stackrel{\text{ind}}{\sim} N(0,\sigma_{\beta_j}^2) \\ \sigma_{\beta_j}^2 &\stackrel{\text{iid}}{\sim} Inverse - Gamma(c,d) \\ \gamma_j &\stackrel{\text{iid}}{\sim} Bernoulli(\rho) \end{split}$$

- $\bullet$  Z is n x p design matrix
- $\Gamma = diag(\gamma_j)$  for  $j = 1, \dots, p$
- a, b, c, d is flat prior

#### 2.2 posterior

$$\begin{split} p(\beta, \sigma^2, \sigma_{\beta}^2, \Gamma | Y) &\propto p(\beta, \sigma^2, \sigma_{\beta}^2, \Gamma, Y) \\ &\propto p(Y | \beta, \sigma^2, \Gamma) p(\beta | \sigma_{\beta}^2) p(\Gamma) p(\sigma^2) p(\sigma_{\beta}^2) \\ &\propto \left(\sigma^2\right)^{-n/2} \exp\left(-\frac{1}{2\sigma^2} \left(Y - Z\Gamma\beta\right)' \left(Y - Z\Gamma\beta\right)\right) \\ &\times \prod_{j=1}^p \left(\sigma_{\beta_j}^2\right)^{-1/2} \exp\left(-\frac{1}{2} \sum_{j=1}^p \frac{\beta_j^2}{\sigma_{\beta_j}^2}\right) \\ &\times \prod_{j=1}^p \rho^{\gamma_j} \left(1 - \rho\right)^{-\gamma_j} \\ &\times \left(\sigma^2\right)^{-a-1} \exp\left(-\frac{b}{\sigma^2}\right) \\ &\times \prod_{j=1}^p \left(\sigma_{\beta_j}^2\right)^{-c-1} \exp\left(-\sum_{j=1}^p \frac{d}{\sigma_{\beta_j}^2}\right) \end{split}$$

#### 2.3 Variational Inference

$$p(\beta, \sigma^2, \sigma_{\beta}^2, \Gamma | Y) \approx q(\beta, \sigma^2, \sigma_{\beta}^2, \Gamma) = \prod_{j=1}^p q_1^*(\beta_j) q_2^*(\sigma^2) \prod_{j=1}^p q_3^*(\sigma_{\beta_j}^2) \prod_{j=1}^p q_4^*(\gamma_j)$$

Use coordinate ascent algorithm, q density of  $\beta$  is

$$q_{1}^{*}(\beta) \propto E_{-q_{1}} \left[ p(\beta, \sigma^{2}, \sigma_{\beta}^{2}, \Gamma, Y) \right]$$

$$\propto E_{-q_{1}} \left[ \exp \left( -\frac{1}{2\sigma^{2}} \left( Y - Z\Gamma\beta \right)' \left( Y - Z\Gamma\beta \right) - \frac{1}{2} \sum_{j=1}^{p} \frac{\beta_{j}^{2}}{\sigma_{\beta_{j}}^{2}} \right) \right]$$

$$\propto \exp \left( -\frac{1}{2} \beta' \left\langle D \right\rangle - \frac{1}{2} \left\langle \frac{1}{\sigma^{2}} \right\rangle \left( \beta' \left\langle \Gamma \right\rangle' Z'Z \left\langle \Gamma \right\rangle \beta - 2\beta' \left\langle \Gamma \right\rangle' Z'Y \right) \right)$$

$$\propto \exp \left( -\frac{1}{2} \left[ \beta' \left( \left\langle D \right\rangle + \left\langle \frac{1}{\sigma^{2}} \right\rangle \left\langle \Gamma \right\rangle' Z'Z \left\langle \Gamma \right\rangle \right) \beta - 2 \left\langle \frac{1}{\sigma^{2}} \right\rangle \beta' \left\langle \Gamma \right\rangle' Z'Y \right] \right)$$

$$\sim N(\mu, \Sigma)$$

Where  $D = diag(\frac{1}{\sigma_{\beta_i}^2})$ , it means expectation under q functions and

$$\Sigma = \left( \langle D \rangle + \beta' \left\langle \frac{1}{\sigma^2} \right\rangle \langle \Gamma \rangle' Z' Z \langle \Gamma \rangle \right)^{-1}, \quad \mu = \left\langle \frac{1}{\sigma^2} \right\rangle \Sigma \langle \Gamma \rangle' Z' Y$$

q density of  $\sigma^2$  is

$$\begin{split} q_{2}^{*}(\sigma^{2}) &\propto E_{-q_{2}}\left[p(\beta,\sigma^{2},\sigma_{\beta}^{2},\Gamma,Y)\right] \\ &\propto E_{-q_{2}}\left[\left(\sigma^{2}\right)^{-\left(\frac{n}{2}+a\right)-1}\exp\left(-\frac{1}{\sigma^{2}}\left(b+\left(Y-Z\Gamma\beta\right)'\left(Y-Z\Gamma\beta\right)\right)\right)\right] \\ &\propto Inverse-Gamma\left(a+\frac{n}{2},b+\frac{1}{2}\left(Y'Y-2\left\langle\beta\right\rangle'\left\langle\Gamma\right\rangle'Z'Y+tr\left(\left(Z'Z\odot\Omega\right)\left(\mu\mu'+\Sigma\right)\right)\right)\right) \end{split}$$

Where odot is hadamard product and

• 
$$\gamma = (\gamma_1, \dots, \gamma_p)$$

• 
$$\Omega = \langle \gamma \rangle \langle \gamma \rangle' + \langle \Gamma \rangle \odot (I - \langle \Gamma \rangle)$$

q density of  $\sigma_{\beta_j}^2$  is

$$q_3^*(\sigma_{\beta}^2) \propto E_{-q_3} \left[ p(\beta, \sigma^2, \sigma_{\beta}^2, \Gamma, Y) \right]$$

$$\propto \prod_{j=1}^p \left[ \left( \sigma_{\beta_j}^2 \right)^{-\left(\frac{1}{2} + c\right) - 1} \exp\left( -\frac{1}{\sigma_{\beta_j}^2} \left( d + \frac{1}{2} \left\langle \beta_j^2 \right\rangle \right) \right) \right]$$

$$q_3^*(\sigma_{\beta_j}^2) \sim Inverse - Gamma \left( c + \frac{1}{2}, d + \frac{1}{2} \left\langle \beta_j^2 \right\rangle \right)$$

q density of  $\gamma$  is

$$q_{4}^{*}(\gamma)E_{-q_{4}}\left[\propto \prod_{j=1}^{p}\rho^{\gamma_{j}}(1-\rho)^{-\gamma_{j}}\exp(-\frac{1}{2}\frac{1}{\sigma^{2}}\left(\beta'\Gamma'Z'Z\Gamma\beta-2\beta'\Gamma'Z'y\right))\right]$$

$$\propto \exp\left[\gamma_{j}\left(\log\left(\rho/(1-\rho)\right)-\frac{1}{2}\left\langle\frac{1}{\sigma^{2}}\right\rangle\left\langle\beta_{j}^{2}\right\rangle Z'_{j}Z_{j}+\left\langle\frac{1}{\sigma^{2}}\right\rangle Z'_{j}\left[Y\mu_{j}-X_{-j}\left\langle\Gamma_{-j}\right\rangle\left(\mu_{-j}\mu_{j}+\Sigma_{-j,j}\right)\right]\right)\right]$$

Where

- $X_j$  means jth coloumn of X
- $X_{-j}$  means without jth column
- $X_{-i,j}$  means jth column without ith component
- $\mu_j$  is jth component of vector and  $mu_j$  means without jth component