

SSVS in Probit Regression

- ▶ Data augmentation Gibbs sampler described earlier easily adapted
- ▶ Sample from conditional posterior of β_j , for $j = 1, \dots, p$:

$$\pi(\beta_j \mid \boldsymbol{\beta}_{(-j)}, \mathbf{z}, \mathbf{y}, \mathbf{X}) = \hat{p}_j \delta_0(\beta_j) + (1 - \hat{p}_j) N(\beta_j; E_j, V_j),$$

$V_j = (c_j^{-2} + \mathbf{X}_j' \mathbf{X}_j)^{-1}$, $E_j = V_j \mathbf{X}_j' (\mathbf{z} - \mathbf{X}_{(-j)} \boldsymbol{\beta}_{(-j)})$, \mathbf{X}_j = j th column of \mathbf{X} , $\mathbf{X}_{(-j)} = \mathbf{X}$ with j th column excluded, $\boldsymbol{\beta}_{(-j)} = \boldsymbol{\beta}$ with j th element excluded, &

$$\hat{p}_j = \frac{p_{0j}}{p_{0j} + (1 - p_{0j}) \frac{N(0; 0, c_j^2)}{N(0; E_j, V_j)}}$$

is the conditional probability of $\beta_j = 0$ (j th predictor excluded)