

Untitled

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12/14/2019

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Input :  $\vec{Y}, \mathbf{X}$ 
Output: Posterior Samples of Regression Coefficients,  $\vec{\beta}$  and  $\vec{\gamma}$ 
1 Set Number of Samples (Total ( $N_s$ ) and Burn-in)
2  $n$  = Number of Observations
3 Initialize  $\vec{\beta}^{(0)}$  and  $\vec{\gamma}^{(0)}$  as the MLE
4 for  $k = 1$  to  $N_s$  do
5   Sample  $\vec{\gamma}_j$  from  $Uniform(max(max(Z_i : Y_i = j), \gamma_{j-1}^{(k-1)}), min(min(Z_i : Y_i = j + 1), \gamma_{j+1}^{(k-1)}))$ 
6   for  $i = 1$  to  $n$  do
7     Sample  $z_i^{(k)} | \vec{\beta}, \vec{\gamma}, y_i = j$  from  $trunc\mathcal{N}(x_i^T \beta^{(k-1)}, 1, \gamma_{j-1}, \gamma_j)$ 
8   end
9   Set  $\Sigma = (X^T X)^{-1}$ 
10  Set  $\vec{\beta}_Z = \Sigma X^T Z$ 
11  Sample  $\beta^{(k)} | \vec{Z}, \vec{\gamma}, \vec{Y}$  from  $\mathcal{N}(\vec{\beta}_Z^{(k)}, \Sigma)$ 
12 end
```

Algorithm 1: Ordered Multinomial Probit Regression Using Gibbs Sampler with Data Augmentation