# notes 1202

### stephanie

#### December 3, 2013

# Contents

| 1 | Hon | nework                  | 1 |
|---|-----|-------------------------|---|
|   | 1.1 | MCMC                    |   |
|   | 1.2 | Probit MCMC             | - |
|   | 1.3 | $\mathrm{C/C}{++}$      | 2 |
|   |     | HW Kernel               |   |
|   | _   |                         | _ |
| 2 | Tru | ncated Normal Sampling  | 2 |
|   | 2.1 | Rejection Sampling      | 2 |
|   |     | 2.1.1 From Robert(2009) | 6 |
|   |     | 2.1.2 Optimal $\alpha$  | : |

# 1 Homework

#### 1.1 MCMC

```
for (iter in 1(niter+burnin)){
if (use GPU )
  Z = rtruncnormGPU( ) CUDA/kernel
else
  Z = rtruncnormCPU( ) regular R or Pythong
beta = rmvnorm ( )
```

### 1.2 Probit MCMC

$$\begin{aligned} \mathbf{Y}_i \mid \mathbf{Z}_i &= \mathbf{I}_{\{}\{Z_i > 0\} \ \mathbf{Z}_i \mid \boldsymbol{\beta} \sim \mathbf{N}(\mathbf{x}_i{}^T \ \boldsymbol{\beta}, \ 1) \ \boldsymbol{\beta} \sim \mathbf{N}(\boldsymbol{\beta}_0, \ \boldsymbol{\Sigma}_0) \\ \mathbf{P}(\boldsymbol{\beta} \mid \mathbf{Z}, \ \mathbf{y}) &\sim \mathbf{Normal} \ \mathbf{P}(\mathbf{Z}_i \mid \boldsymbol{\beta}, \ \mathbf{y}_i) \sim \mathbf{Truncated normal} \end{aligned}$$

#### 1.3 C/C++

- C is a very fast compiled language
- Data types need to be explicitly defined.
- Vectors/matrics are typically implemented using pointers
- Pointers point to memory locations, from which can look up values at the memory locations.

#### 1.4 HW Kernel

- Use the template in the github repo.
- void says the function doesn't return anything. You want the kernels to be void. put the return values into one of the arguments.
- See notes, they are on pearson example

# 2 Truncated Normal Sampling

#### 2.1 Rejection Sampling

To sample form a distribution with pdf f(x) if we can find another distribution with pdf g(x) such that

$$f(x) \le Mg(x)$$
 for all  $x$  (1)

then we can use g to sample from f as follows

- 1. Sample a value  $x_*$  from g(x)
- 2. Sample  $U \sim U[0,1]$
- 3. If  $U \leq \frac{f(x_*)}{Mg(x_*)}$  then accept  $x_*$ , otherwise, return to 1.

Ideally f(x) and g(x) should be close.

#### 2.1.1 From Robert (2009)

To sample from  $X \sim N(0,1,\mu^-,\infty)$ 

- 1. Generate  $Z = \mu^- + Expo(\alpha)$
- 2. Compute

$$\psi(z) = \begin{cases} \exp(-\frac{(\alpha - z)^2}{2} & if\mu^- < \alpha \\ \exp(-\frac{(\mu^- - \alpha)^2}{2}) \exp(\frac{(\alpha - z)^2}{2}) & if\mu^- \ge \alpha \end{cases}$$
 (2)

If  $U[0,1] < \psi(z)$  accept. Else, try again

#### 2.1.2 Optimal $\alpha$

$$\alpha = \frac{\mu^- + \sqrt{(\mu^-)^2 + 4}}{2} \tag{3}$$

We need  $X \sim N(\mu, \sigma^2; a, \infty)$  Let  $Z \sim N(0, 1; \mu, \infty)$  What is the distribution of Y = cZ + k?

$$Y \sim N(k, c^2, k + c\mu^-, \infty) \tag{4}$$