

STAT6011/7611/6111/3317  
COMPUTATIONAL STATISTICS (2016 Fall)  
Assignment 3

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1.1

$$\begin{aligned} p(\mathbf{y}|\sigma^2) &= \int \cdots \int \prod_{i=1}^n \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(y_i - \theta_i)^2}{2}\right) \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{\theta_i^2}{2\sigma^2}\right) d\theta_1 \cdots d\theta_n \\ &= \int \cdots \int \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sqrt{\sigma^2+1}} \exp\left(-\frac{y_i}{2} + \frac{\sigma^2 y_i^2}{2(\sigma^2+1)}\right) \frac{\sqrt{\sigma^2+1}}{\sqrt{2\pi}\sigma} \exp\left(\frac{\sigma^2+1}{\sigma^2} \left(\theta_i - \frac{\sigma^2}{\sigma^2+1} y_i\right)\right) d\theta_1 \cdots d\theta_n \\ &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sqrt{\sigma^2+1}} \exp\left(-\frac{y_i}{2} + \frac{\sigma^2 y_i^2}{2(\sigma^2+1)}\right) \\ &= \left(\frac{1}{\sqrt{2\pi}\sqrt{\sigma^2+1}}\right)^n \exp\left(-\frac{\sum_{i=1}^n y_i}{2} + \frac{\sigma^2 \sum_{i=1}^n y_i^2}{2(\sigma^2+1)}\right) \end{aligned}$$

1.2

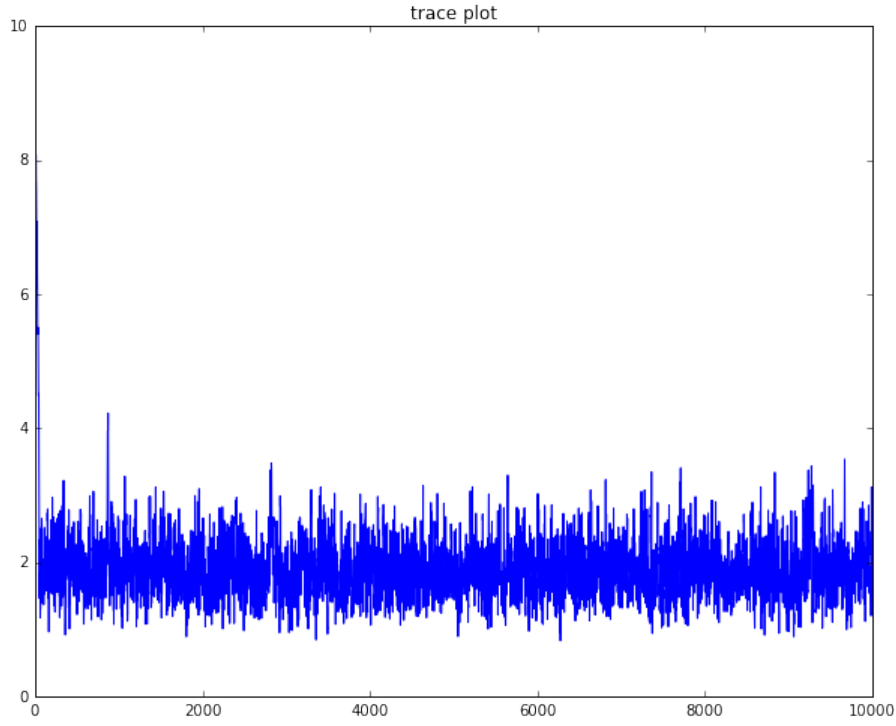
$$p(\sigma^2|\mathbf{y}) \propto \left(\frac{1}{\sqrt{2\pi}\sqrt{\sigma^2+1}}\right)^n \exp\left(-\frac{\sum_{i=1}^n y_i}{2} + \frac{\sigma^2 \sum_{i=1}^n y_i^2}{2(\sigma^2+1)}\right) \frac{1}{\sigma^2}$$

Listing 1: 1-b

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```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 % matplotlib inline
4 import pandas as pd
5
6 y = q1.values[:, 1]
7 def pos(sigma):
8     return np.exp(sigma*np.dot(y, y)/(2*(sigma + 1)) - sum(y)/2) / (sigma*(np.sqrt(2*np.pi * (sigma + 1))))**100
9
10 np.random.seed(1234)
11 samples = [10]
12 for i in range(10000):
13     u = np.random.uniform()
14     propose = samples[-1] + np.random.normal(scale = 0.5)
15     prob = min(1, pos(propose)/pos(samples[-1]))
16
17     if prob < u:
18         samples.append(propose)
19     else:
20         samples.append(samples[-1])
21
22 plt.figure(figsize = (10, 8))
23 plt.title("traceuplot")
24 plt.plot(samples)
25 plt.show()
```

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### 1.3

Let  $\Theta = (\theta_1, \theta_2, \dots, \theta_n)$ , and let  $\Theta_{-j} = (\theta_1, \dots, \theta_{j-1}, \theta_{j+1}, \dots, \theta_n)$ .

$$p(\sigma^2 | \mathbf{y}, \Theta) \propto \frac{1}{\sigma^2 (\sqrt{2\pi\sigma^2})^n} \exp\left(-\frac{\sum_i \theta_i^2}{2\sigma^2}\right)$$

$$p(\theta_j | \mathbf{y}, \Theta_{-j}, \sigma^2) \propto N\left(\frac{\sigma^2 y_j}{\sigma^2 + 1}, \frac{\sigma^2}{\sigma^2 + 1}\right) \forall j \in \{1, 2, 3, \dots, n\}$$

Listing 2: 1-c

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```

1 def pos2(sigma, i):
2     return np.exp(-np.dot(theta_sample[:, i], theta_sample[:, i])/(2*sigma)) / (sigma*(np.sqrt(2*np.pi * sigma))**100)
3
4 # Gibbs sampler
5 np.random.seed(1234)
6
7 sigma_sample = [5]
8 theta_sample = np.ones((len(y), 10000))
9
10 for i in range(10000):
11
12     if i != 0:
13         u = np.random.uniform()
14         propose = sigma_sample[-1] + np.random.normal(scale = 0.2)
15         prob = min(1, pos2(propose, i)/pos2(sigma_sample[-1], i))
16
17         if prob < u:
18             sigma_sample.append(sigma_sample[-1])
19         else:
20             sigma_sample.append(propose)
21
22     for j in range(100):
23         theta_sample[j, i] = np.random.normal(loc = sigma_sample[-1]*y[j]/(sigma_sample[-1] + 1), scale = np.
24             sqrt(sigma_sample[-1]/(sigma_sample[-1] + 1)))
25
26 else:
27     for j in range(100):

```

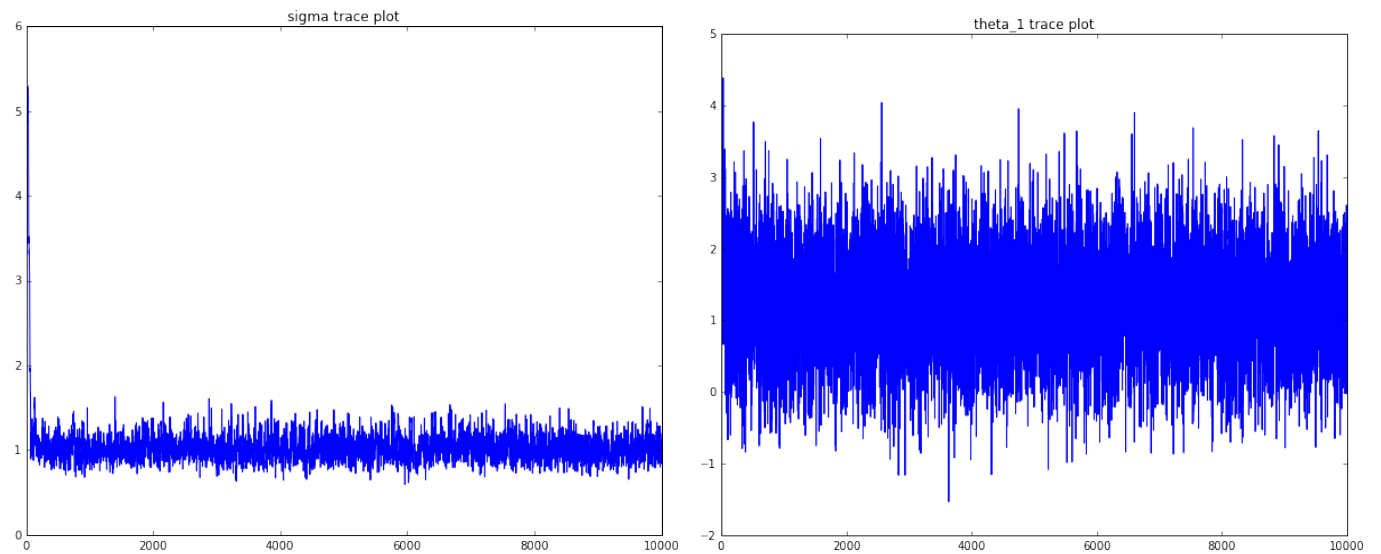
```

27         theta_sample[j, i] = np.random.normal(loc = sigma_sample[-1]*y[j]/(sigma_sample[-1] + 1), scale = np.
          sqrt(sigma_sample[-1]/(sigma_sample[-1] + 1)))
28
29
30 # plot sigma
31 plt.figure(figsize = (10, 8))
32 plt.title("sigma_trace_plot")
33 plt.plot(sigma_sample)
34 plt.xlim((0, 10000))
35 plt.show()
36
37 # plot first 3 theta
38 for i in range(3):
39     plt.figure(figsize = (10, 8))
40     plt.title("theta_1_trace_plot")
41     plt.plot(theta_sample[i, :])
42     plt.xlim((0, 10000))
43     plt.show()

```

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Trace plot.  $\sigma^2$  and  $\theta_1$ .



- 2
- 2.1
- 2.2
- 2.3
- 2.4
- 3
- 3.1
- 3.2