### **VE414 Bayesian Analysis**

## Fall 2022 - Homework3 Multiparameter Model



This homework is done by Wei Linda

### 1 Problem 1

(a)

Since We don't know the variance, we shall use T distribution:

$$T_{n-1} = \frac{\bar{x} - \theta}{s / \sqrt{n}}$$

follows a T distribution, so the 95% confidence interval for  $\theta$  is  $\bar{x} \pm t_{0.025,n-1} \cdot S/\sqrt{n} = (7.97822, 9.11842)$ 

(b)

$$f(n, \sigma^{2} | x)$$

$$\propto f(x | n, \sigma^{2}) \cdot \pi(n, \sigma^{2})$$

$$\propto \sigma^{-2} \prod_{i=1}^{14} \frac{1}{\sqrt{2\pi}\sigma} \cdot e^{-\frac{1}{2} \frac{(x_{i} - \theta)^{2}}{\sigma^{2}}}$$

$$\propto \sigma^{-2-n} e^{-\frac{1}{2\sigma^{2}} \left(\sum_{i=1}^{14} (x_{i} - \theta)^{2}\right)}$$

$$\propto \sigma^{-2-n} e^{-\frac{1}{2\sigma^{2}} \left(\sum_{i=1}^{14} ((n-1)S^{2} + n(\bar{x} - \theta)^{2})\right)}$$

$$p(\theta \mid x) = \int_0^\infty p(\theta, \sigma^2 \mid x) d\sigma^2$$

$$\propto \int_0^\infty \sigma^{-2-n} \cdot e^{-\frac{1}{2\sigma^2} \left[ (n-1)s^2 + n(\bar{x}-n)^2 \right]}$$
$$= t_{n-1}(\theta|\bar{x}, s^2/n)$$

That is to say

$$\frac{\bar{x} - \theta}{s / \sqrt{n}} \sim T_{n-1}$$

Thus, 95% credible interval

$$\bar{x} \pm t_{0.025.13} S /= (7.97822, 9.11842)$$

which is the same like (a).

(c) According to the BDA3 textbook.

$$p(n \mid y) \propto \left(1 + \frac{k_n(\theta - \theta n)^2}{V_n^2 \sigma_n^2}\right)^{-\frac{V_{n+1}}{2}}$$

$$= t_{vn} \left( \mu_n, \sigma_n^2 / k_n \right)$$

$$\mu_n = \frac{k_0}{k_0 + n} n_0 + \frac{n}{k_0 + n} \bar{x}$$

$$= \frac{10}{10 + 14} 8.35 + \frac{14}{10 + 14} 8.55$$

$$= 8.4867$$

$$k_n = 24$$

$$V_n = 4 + 14 = 18$$

$$18\sigma_n^2 = 4 \times 1.5 + (14 - 1) \times 0.9758 + \frac{10 \times 14}{10 + 14} \times (8.55 - 8.35)^2$$

$$\sigma_n^2 = 1.05$$

$$P(\theta \mid x) = t_{18} \left( 8.4667, \sqrt{\frac{1.05}{24}} \right)$$
$$\frac{\mu_n - 8.4667}{\sqrt{\frac{1.05^2}{24}}} \sim T_{18}.$$

- the 95% credible interval is  $8.4667 \pm t_{18,0.025} \times \sqrt{\frac{1.05}{24}} = (8.02, 8.91)$ 

#### 2 Problem 2

$$f(U) = \begin{cases} 1 & u \in (0,1) \\ 0 & \text{else} \end{cases}$$
$$F(U) = \begin{cases} 0 & u \in (-\infty,0) \\ u & u \in (0,1) \\ 1 & u \ge 1 \end{cases}$$

 $F[U \le u] = u. \ F[g(u) \le u] = FFU \le g^{-1}(u)I = 1 - \frac{25}{u^2} = g^{-1}(u) \Rightarrow g(u) = \sqrt{\frac{25}{1-u}} \text{ propose a transformation}$ 

$$x = g(u) = \sqrt{\frac{25}{1 - u}}$$

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When *n*= 0.36

$$x = \frac{25}{4}$$

#### 3 Problem 3

$$P(\theta \mid x) = P(x \mid \theta) \cdot \pi(\theta)$$

$$= \prod_{i=1}^{5} p(x_i \mid \theta) \pi(\theta)$$

$$\propto \frac{2^4}{T(4)} \cdot \theta^3 e^{-2\theta} \prod_{i=1}^{5} \theta \cdot e^{-\theta x_i}$$

$$\propto \theta^8 e^{-(2+5\bar{x})\theta}.$$

$$\sim \Gamma(9, 2+5\bar{x})$$

$$\sim \Gamma(9, 10)$$

So, mean is  $E = \frac{9}{10} = 0.9$  and Variance is 0.09.

# 4 Problem 4

```
In [3]: import numpy as np
    from scipy.stats import uniform
    import seaborn as sns
    import matplotlib.pyplot as plt
    u = uniform(0, 1).rvs(10000)
    x=-np.log(1-u)/20
    sns.histplot(x)
    plt.title('Exponential distribution simulation')
```

Out[3]: Text(0.5, 1.0, 'Exponential distribution simulation')

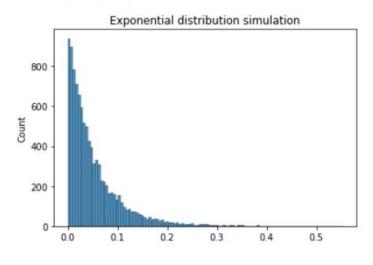


图 1: Problem 4

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