

**Problem 1**

Answer to the problem goes here.

1. Problem 1 part 1 answer here.

$$\begin{aligned} f(\theta|y) &\propto f(y|\theta)f(\theta) = \prod_{i=1}^n \sqrt{\frac{2}{\pi}} \theta^{3/2} y_i^2 e^{-\frac{\theta y_i^2}{2}} \lambda e^{-\lambda \theta} \propto \theta^{\frac{3n}{2}} e^{-\lambda \theta - \frac{\theta}{2} \sum_{i=1}^n y_i^2} \\ &= \theta^{\frac{3n}{2}+1-1} e^{-\theta(\lambda + \frac{1}{2} \sum_{i=1}^n y_i^2)} \end{aligned}$$

Which is the pdf of Gamma distribution  $(\frac{3n}{2} + 1, \lambda + \frac{1}{2} \sum_{i=1}^n y_i^2)$

2. Problem 1 part 2 answer here.

The posterior mean is  $\frac{\frac{3n}{2}+1}{\lambda + \frac{1}{2} \sum_{i=1}^n y_i^2}$ . By plugging the number,  $\frac{\frac{3*3}{2}+1}{0.5 + \frac{1}{2}(1.4^2 + 3.1^2 + 2.5^2)} = 0.5844846$

The MLE is  $\frac{3n}{\sum_{i=1}^n y_i^2} = \frac{9}{1.4^2 + 3.1^2 + 2.5^2} = 0.5050505$

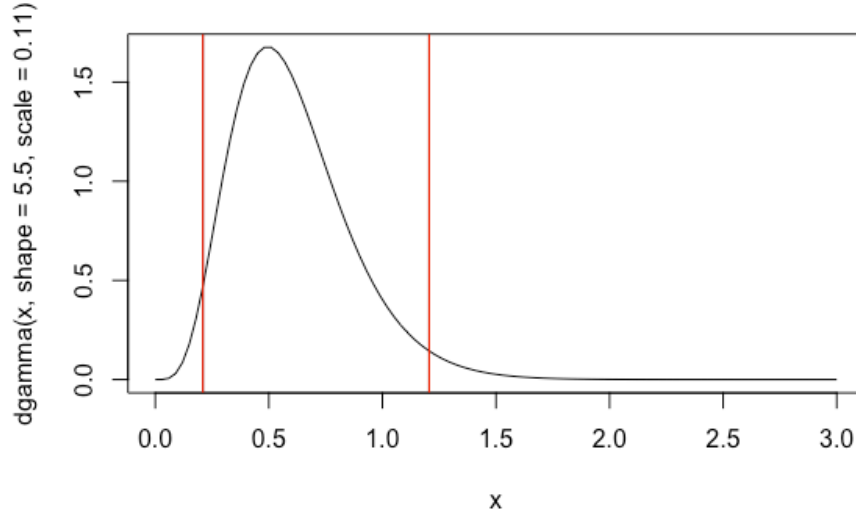
The prior mean is  $\frac{1}{\lambda} = 2$

3. Problem 1 part 3 answer here

R code:

```
plot(0, 0, xlim = c(0, 3), ylim = c(0, 1), type = "n")
curve(dgamma(x, shape = 5.5, scale = 0.11), from = 0, to = 3)
q1<-qgamma(0.025,shape = 5.5, scale = 0.11)
q2<-qgamma(0.975,shape = 5.5, scale = 0.11)
abline(v=c(q1,q2),col="red")
```

The 95% equitailed credible set is [0.2098662, 1.205603]



4. Problem 1 part 4 answer here

Gamma distribution  $(\frac{3n}{2} + 1, \lambda + \frac{1}{2}\sum_{i=1}^n y_i^2)$ , posterior gamma (5.5, 9.41)

$$\text{As } E(Y) = 2\sqrt{\frac{2}{\pi\theta}}$$

$$\begin{aligned}\hat{y}_{n+1} &= \int_0^\infty 2\sqrt{\frac{2}{\pi\theta}} \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta} d\theta = \int_0^\infty 2\sqrt{\frac{2}{\pi\theta}} \frac{9.41^{5.5}}{\Gamma(5.5)} \theta^{5.5-1} e^{-9.41\theta} d\theta \\ &= \left(\frac{9.41^{5.5}}{\Gamma(5.5)}\right) \frac{\Gamma(5)}{9.41^5} 2\sqrt{\frac{2}{\pi}} \int_0^\infty \frac{9.41^5}{\Gamma(5)} \theta^{5-1} e^{-9.41\theta} d\theta = 2.244499\end{aligned}$$

## Problem 2

Answer to the problem goes here.

1. Problem 2 part 1 answer here.

$$\begin{aligned}\pi(\theta|x) &= \frac{f(x|\theta)\pi(\theta)}{m(x)} = \frac{f(x|\theta)[\epsilon\pi_1(\theta) + (1-\epsilon)\pi_2(\theta)]}{\int_{\Theta} f(x|\theta)[\epsilon\pi_1(\theta) + (1-\epsilon)\pi_2(\theta)]d\theta} \\ &= \frac{\epsilon f(x|\theta)\pi_1(\theta) + (1-\epsilon)\pi_2(\theta)f(x|\theta)}{\epsilon \int_{\Theta} f(x|\theta)\pi_1(\theta)d\theta + (1-\epsilon) \int_{\Theta} f(x|\theta)\pi_2(\theta)d\theta} \\ &= \frac{\epsilon\pi_1(\theta|x)m_1(\theta) + (1-\epsilon)\pi_2(\theta|x)m_2(x)}{\epsilon m_1(\theta) + (1-\epsilon)m_2(x)} \\ &= \epsilon'\pi_1(\theta|x) + (1-\epsilon')\pi_2(\theta|x)\end{aligned}$$

2. Problem 2 part 2 answer here:

$$\begin{aligned}\pi(\theta|x) &\propto f(x|\theta)\pi(\theta) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\theta)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi\sigma_0^2}} e^{-\frac{(\theta-\theta_0)^2}{2\sigma_0^2}} \\ &\quad \frac{\theta^2 - \frac{2(\sigma^2\theta_0 + \sigma_0^2x)}{\sigma_0^2 + \sigma^2}\theta + \frac{\sigma^2\theta_0^2 + \sigma_0^2x^2}{\sigma_0^2 + \sigma^2}}{-\frac{2\sigma^2\sigma_0^2}{\sigma_0^2 + \sigma^2}} \\ &= \frac{1}{2\pi\sigma\sigma_0} e^{\frac{\sigma^2\theta_0 + \sigma_0^2x}{\sigma_0^2 + \sigma^2} - \frac{\sigma^2\sigma_0^2}{\sigma_0^2 + \sigma^2}}\end{aligned}$$

Which could be represented by  $N(\frac{\sigma^2\theta_0 + \sigma_0^2x}{\sigma_0^2 + \sigma^2}, \frac{\sigma^2\sigma_0^2}{\sigma_0^2 + \sigma^2})$

Therefore,  $\pi_1(\theta) \sim N(\frac{60*98 + 80*110}{80+60}, \frac{60*80}{80+60})$ , i.e.  $N(104.9, 34.3)$

$\pi_2(\theta) \sim N(\frac{200*98 + 80*100}{80+200}, \frac{200*80}{80+200})$ , i.e.  $N(98.6, 57.1)$

$$\begin{aligned}m(x) &= \int_{-\infty}^{\infty} f(x|\theta)\pi(\theta)d\theta \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\theta)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi\sigma_0^2}} e^{-\frac{(\theta-\theta_0)^2}{2\sigma_0^2}} d\theta \\ &= \frac{\sqrt{2\pi\frac{\sigma^2\sigma_0^2}{\sigma_0^2 + \sigma^2}}}{2\pi\sigma\sigma_0} e^{\frac{(\frac{\sigma^2\theta_0 + \sigma_0^2x}{\sigma_0^2 + \sigma^2})^2 - \frac{\sigma^2\theta_0^2 + \sigma_0^2x^2}{\sigma_0^2 + \sigma^2}}{2\frac{\sigma^2\sigma_0^2}{\sigma_0^2 + \sigma^2}}}\end{aligned}$$

As  $\pi_1(\theta) \sim N(110, 60)$ ,  $m_1(98) = 0.0202$

As  $\pi_2(\theta) \sim N(100, 200)$ ,  $m_2(98) = 0.0237$

$$\epsilon' = \frac{\frac{2}{3}m_1(98)}{\frac{2}{3}m_1(98) + \frac{1}{3}m_2(98)} = 0.63$$

Then the posterior is  $0.63 * N(104.9, 34.3) + 0.37 * N(98.6, 57.1)$

The bayes estimator is expectation of this mixture distribution  $0.63*104.9 + 0.37*98.6 = 102.6$

### Problem 3

Answer to the problem goes here.

3. Problem 3(a) part 1 answer here

Beta prior mean:  $\frac{\alpha}{\alpha+\beta} = \frac{15}{20} = 0.75$

Posterior distribution: Beta (15+787, 5+1064-787), i.e. Beta(802, 282)

The posterior mean is  $802/(802+282)=0.7398524$

4. Problem 3(a) part 2 answer here:

R code: `pbeta(0.75,802,282)`

$$P(p \leq 3/4) = 0.7754435$$

5. Problem 3(a) part 3 answer here:

R code:

`qbeta(0.025,802,282): 0.7133363`

`qbeta(0.975,799,281): 0.7655302`