

Math 170S HW5

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1) a) Posterior pdf of θ :

Since Y is the sum of n observations,

Y is a Poisson distr. with $\lambda = n\theta$

$$\text{Given } Y=y: g(y|\theta) = \frac{n^y \theta^y e^{-n\theta}}{y!}$$

The prior pdf is gamma distr. with params α & β :

$$h(\theta) = \frac{\theta^{\alpha-1} e^{-\theta/\beta}}{\Gamma(\alpha) \cdot \beta^\alpha}$$

So, our posterior pdf $K(\theta|y)$ is $\frac{g(y|\theta)h(\theta)}{\int_{-\infty}^{\infty} g(y|\theta)h(\theta)d\theta}$

We evaluate the integral first:

$$\int_{-\infty}^{\infty} \frac{n^y \theta^y e^{-n\theta}}{y!} \cdot \frac{\theta^{\alpha-1} e^{-\theta/\beta}}{\Gamma(\alpha) \cdot \beta^\alpha} d\theta$$

Take out all constants then simplify:

$$\frac{n^y}{y! \Gamma(\alpha) \beta^\alpha} \int_0^{\infty} \theta^y e^{-n\theta} \theta^{\alpha-1} e^{-\theta/\beta} d\theta = \frac{n^y}{y! \Gamma(\alpha) \beta^\alpha} \int_0^{\infty} \theta^{y+\alpha-1} e^{-\theta(n+\frac{1}{\beta})} d\theta$$

Turn the integrand into a gamma distr:

$$= \frac{n^y \overline{\Gamma_{y+d}} \left(\frac{1}{n+\frac{1}{\beta}}\right)^{y+d}}{y! \overline{\Gamma_d} \beta^d} \int_0^\infty \frac{\theta^{y+d-1} e^{-\theta(n+\frac{1}{\beta})}}{\overline{\Gamma_{y+d}} \left(\frac{1}{n+\frac{1}{\beta}}\right)^{y+d}} d\theta$$

$\underbrace{\hspace{10em}}_{\text{gamma}\left(y+d, \frac{1}{n+\frac{1}{\beta}}\right)}$

so this resolves to 1

$$= \frac{n^y \overline{\Gamma_{y+d}} \beta^{y+d}}{y! \overline{\Gamma_d} \beta^d (n\beta+1)^{y+d}} = \frac{n^y \overline{\Gamma_{y+d}} \beta^y}{y! \overline{\Gamma_d} (n\beta+1)^{y+d}}$$

Now, the numerator:

$$g(y|\theta)h(\theta) = \frac{n^y \theta^y e^{-n\theta}}{y!} \cdot \frac{\theta^{d-1} e^{-\theta/\beta}}{\overline{\Gamma_d} \beta^d}$$

Combining:

$$\frac{\cancel{n^y} \theta^y e^{-n\theta}}{y!} \cdot \frac{\theta^{d-1} e^{-\theta/\beta}}{\cancel{\overline{\Gamma_d}} \beta^d} = \frac{\theta^{y+d-1} e^{-n\theta - \theta/\beta} (n\beta+1)^{y+d}}{\overline{\Gamma_{y+d}} \beta^{d+y}}$$

$$\frac{\cancel{n^y} \overline{\Gamma_{y+d}} \beta^y}{y! \cancel{\overline{\Gamma_d}} (n\beta+1)^{y+d}} = \frac{\theta^{y+d-1} e^{-n\theta - \theta/\beta}}{\overline{\Gamma_{y+d}} \left(\frac{1}{n+\frac{1}{\beta}}\right)^{y+d}}$$

So, this gives us a gamma distribution
with parameters $(\alpha + y, \frac{1}{n + \frac{1}{\beta}})$

b) The loss function is minimized when $w(y)$ is the mean of our posterior pdf. The mean of a gamma distribution is simply the product of the 2 parameters. So,

$$w(y) = E(Y|\theta) = \frac{\alpha + y}{n + \frac{1}{\beta}}$$

c) Now, the above Bayes' Estimator can be rewritten as:

$$\frac{\alpha}{n + \frac{1}{\beta}} + \frac{y}{n + \frac{1}{\beta}} = \left(\frac{\frac{1}{\beta}}{n + \frac{1}{\beta}} \right) \cdot \alpha\beta + \left(\frac{n}{n + \frac{1}{\beta}} \right) \cdot \frac{y}{n}$$

Thus, this shows $w(y)$ is a weighted avg. of MLE $\frac{y}{n}$ and prior mean $\alpha\beta$, with respective weights as circled above.

2) a) X_1, \dots, X_n has a gamma distr.

with params α known and $\theta = \frac{1}{\tau}$

$$\text{We have } g(x_i | \tau) = \frac{x_i^{\alpha-1} e^{-\tau x_i}}{\Gamma(\alpha) \left(\frac{1}{\tau}\right)^\alpha}$$

$$\begin{aligned} g(x_1, \dots, x_n | \tau) &= g(x_1 | \tau) \cdot g(x_2 | \tau) \cdots g(x_n | \tau) \\ &= \frac{\left(\prod_{i=1}^n x_i\right)^{\alpha-1} e^{-\tau \sum_{i=1}^n x_i}}{\left(\Gamma(\alpha) \left(\frac{1}{\tau}\right)^\alpha\right)^n} \end{aligned}$$

Our prior pdf $h(\tau)$ is a gamma distr. with
params α_0 & $\theta_0 \Rightarrow h(\tau) = \frac{\tau^{\alpha_0-1} e^{-\tau/\theta_0}}{\Gamma(\alpha_0) \theta_0^{\alpha_0}}$

Similar to Q1, we will calculate the integral:

$$\int_{-\infty}^{\infty} g(x | \tau) h(\tau) d\tau = \int_0^{\infty} \frac{\left(\prod_{i=1}^n x_i\right)^{\alpha-1} e^{-\tau \sum_{i=1}^n x_i}}{\left(\Gamma(\alpha) \left(\frac{1}{\tau}\right)^\alpha\right)^n} \cdot \frac{\tau^{\alpha_0-1} e^{-\tau/\theta_0}}{\Gamma(\alpha_0) \theta_0^{\alpha_0}} d\tau$$

Take out constants:

$$\frac{\left(\prod_{i=1}^n x_i\right)^{\alpha-1}}{\Gamma(\alpha)^n \Gamma(\alpha_0) \theta_0^{\alpha_0}} \int_0^{\infty} \tau^{\alpha n} e^{-\tau \sum_{i=1}^n x_i} \tau^{\alpha_0-1} e^{-\tau/\theta_0} d\tau$$

Simplify:

$$\frac{\left(\prod_{i=1}^n x_i\right)^{d-1}}{\Gamma(d)^n \Gamma(d_0) \theta_0^{d_0}} \int_0^\infty \tau^{dn+d_0-1} e^{-\tau \left(\sum_{i=1}^n x_i + \frac{1}{\theta_0}\right)} d\tau$$

We again use the technique of turning the integrand into a gamma distr. and we are left with:

$$\frac{\left(\prod_{i=1}^n x_i\right)^{d-1} \Gamma(dn+d_0) \left(\frac{1}{\sum_{i=1}^n x_i + \frac{1}{\theta_0}}\right)^{dn+d_0}}{\Gamma(d)^n \Gamma(d_0) \theta_0^{d_0}}$$

$$= \frac{\left(\prod_{i=1}^n x_i\right)^{d-1} \Gamma(dn+d_0) \theta_0^{nd}}{\Gamma(d)^n \Gamma(d_0) \left(\sum_{i=1}^n \theta_0 x_i + 1\right)^{dn+d_0}}$$

Combining all:

$$\frac{\cancel{\left(\prod_{i=1}^n x_i\right)^{d-1}} e^{-\tau \sum_{i=1}^n x_i} \tau^{d_0-1} e^{-\tau/\theta_0}}{\cancel{\left(\Gamma(d) \left(\frac{1}{\tau}\right)^d\right)^n} \Gamma(d_0) \theta_0^{d_0}}$$

$$\frac{\cancel{\left(\prod_{i=1}^n x_i\right)^{d-1}} \Gamma(dn+d_0) \theta_0^{nd}}{\cancel{\Gamma(d)^n \Gamma(d_0)} \left(\sum_{i=1}^n \theta_0 x_i + 1\right)^{dn+d_0}}$$

$$\begin{aligned}
 &= \frac{\tau^{\alpha n} \cdot \tau^{\alpha_0 - 1} \cdot e^{-\tau \sum_{i=1}^n x_i} \cdot e^{-\tau/\theta_0} \cdot \left(\sum_{i=1}^n \theta_0 x_i + 1 \right)^{\alpha n + \alpha_0}}{\Gamma(\alpha n + \alpha_0) \cdot \theta_0^{\alpha n} \cdot \theta_0^{\alpha_0}} \\
 &= \frac{\tau^{\alpha n + \alpha_0 - 1} e^{-\tau \left(\sum_{i=1}^n x_i + \frac{1}{\theta_0} \right)}}{\Gamma(\alpha n + \alpha_0) \cdot \left(\sum_{i=1}^n x_i + \frac{1}{\theta_0} \right)^{\alpha n + \alpha_0}}
 \end{aligned}$$

This posterior pdf is a gamma distribution with parameters $\alpha n + \alpha_0$ and $\frac{1}{\sum_{i=1}^n x_i + \frac{1}{\theta_0}}$

b) The mean of the above gamma distribution is the product of the 2 parameters:

$$\frac{\alpha n + \alpha_0}{\sum_{i=1}^n x_i + \frac{1}{\theta_0}} = \frac{\alpha n + \alpha_0}{n\bar{X} + \frac{1}{\theta_0}} = \boxed{\frac{\alpha n \theta_0 + \alpha_0 \theta_0}{n\theta_0 \bar{X} + 1}}$$

The boxed mean is written in terms of sample mean \bar{X} and prior mean $\alpha_0 \theta_0$. ✓

3) We realize that through Q1 and Q2 that we don't actually need to calculate $K_1(y)$ (the integral) because a lot of things end up canceling out in the end. For this question, we will use the concept of proportionality (\propto) to simplify our problem:

So, we have from given that our prior distr:

$$\pi(\theta) \propto \theta^{\alpha-1} e^{-\beta\theta}, \theta > 0$$

and $y \sim \text{Exp}(\theta)$ with $y = \{y_1, \dots, y_n\}$ iid

$$\text{so } g(y_i | \theta) \propto \theta e^{-y_i \theta}$$

$$\begin{aligned} \Rightarrow g(y | \theta) &= g(y_1 | \theta) \cdot g(y_2 | \theta) \cdots g(y_n | \theta) \\ &\propto \theta^n e^{-(y_1 + y_2 + \cdots + y_n)\theta} \end{aligned}$$

Our posterior pdf:

$$\begin{aligned} k(\theta | y) &\propto \theta^{\alpha-1} e^{-\beta\theta} \cdot \theta^n e^{-(y_1 + y_2 + \cdots + y_n)\theta} \\ &\propto \theta^{\alpha+n-1} e^{-\theta(\beta + n\bar{y})} \end{aligned}$$

Based on this proportion, our posterior pdf is a gamma distribution with parameters $(\alpha+n, \beta+n\bar{y})$

