

1.

Our Jeffrey's prior is the following:

$$\pi(\theta) = \frac{1}{\sqrt{\theta}}$$

And our observations:  $x_1 = 1, x_2 = 2, x_3 = 0, x_4 = 1$

Since we're dealing with Poisson we know our likelihood function is in the following form:

$$L(\theta|x) = \frac{\theta^{x_i} e^{-\theta}}{x_i!}$$

When inputting these values we get the following for the likelihood function:

$$= e^{-4\theta} \left( \frac{\theta^1}{1} * \frac{\theta^2}{2} * \frac{\theta^0}{1} * \frac{\theta^1}{1} \right)$$

$$= \theta^4 e^{-4\theta}$$

$$L(\theta|x) * \pi(\theta) = \theta^{\frac{7}{2}} e^{-4\theta}$$

Now that we have the result we can use this and plug it into the bayes estimator formula:

$$\delta_B(x) = \frac{\int \theta^{\frac{9}{2}} e^{-4\theta} d\theta}{\int \theta^{\frac{7}{2}} e^{-4\theta} d\theta}$$

Now since we know we can introduce the gamma function and have that integrate to 1 in order to simplify:

$$\alpha = \frac{11}{2}$$

$$\beta = 4$$

$$\begin{aligned}
& \frac{\Gamma\left(\frac{11}{2}\right)}{4^{\frac{11}{2}}} \int \frac{4^{\frac{11}{2}}}{\Gamma\left(\frac{11}{2}\right)} \theta^{\frac{11}{2}-1} e^{-4\theta} d\theta \\
&= \frac{\Gamma\left(\frac{9}{2}\right)}{4^{\frac{9}{2}}} \int \frac{4^{\frac{9}{2}}}{\Gamma\left(\frac{9}{2}\right)} \theta^{\frac{9}{2}-1} e^{-4\theta} d\theta \\
&= \frac{\Gamma\left(\frac{11}{2}\right) * 4^{\frac{11}{2}}}{\Gamma\left(\frac{9}{2}\right) * 4^{\frac{9}{2}}} \\
&= 9/2 * 1/4 \\
&= \frac{9}{8}
\end{aligned}$$

Which is very close to 1!

b)

Credible set can be derived using the available function in R:

Where the shape value is our alpha = 11/2 - 1 = 4.5 and our rate value which represents Beta = 4.

```

> {r cars}
> qqgamma(c(0.025,0.975),4.5,4)
[1] 0.3375487 2.3778460

```

c)

Same can be done for the HDP credible set and we get a similar result:

```

1
2 ▾ ## Including Plots
3
4 You can also embed plots, for example:
5
6 ▾ ```{r pressure, echo=FALSE}
7 library(HDIinterval)
8 int <- hdi(qgamma,shape=4.5,rate=4)
9 print(int)
10
11 ▴ ```

```

```

      lower      upper
0.2378233 2.1740332
attr(,"credMass")
[1] 0.95

```

d)

From the derivation we know that the MAP estimator for poisson is the following:

$$\begin{aligned}
 & \arg \max_{\lambda} l(\lambda) \\
 & \frac{\partial}{\partial \lambda} l(\lambda) = 0 \\
 & \frac{\partial}{\partial \lambda} \left\{ \sum_{i=1}^n \ln(e^{-\lambda}) + \sum_{i=1}^n x_i \ln \lambda - \sum_{i=1}^n \ln x_i! \right\} = 0 \\
 & -n + \frac{1}{\lambda} \sum_{i=1}^n x_i = 0 \\
 & \therefore \hat{\lambda} = \frac{1}{n} \sum_{i=1}^n x_i
 \end{aligned}$$

Therefore, it equals  $= 1/4 (4) = 1$