

Homework 3

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1.a)

$$\mu = 0.95, \sigma = 0.05 \Rightarrow 0.95 = \frac{a}{a+b}, 0.05^2 = \frac{ab}{(a+b)^2(a+b+1)}$$

$$\Downarrow$$

$$19b = a, 0.05^2 = \frac{19b^2}{(20b)^2(20b+1)}$$

$$\Downarrow$$

Expert 1 $\sim \text{Beta}(17.1, 0.9)$ prior

$$b = 0.9, a = 17.1$$

$$\mu = 0.5, \sigma = \frac{1}{12} \Rightarrow 0.5 = \frac{a}{a+b}, \left(\frac{1}{12}\right)^2 = \frac{ab}{(a+b)^2(a+b+1)}$$

$$\Downarrow$$

$$b = a, \left(\frac{1}{12}\right)^2 = \frac{b^2}{(2b)^2(2b+1)}$$

$$\Downarrow$$

Expert 2 $\sim \text{Beta}(17.5, 17.5)$ prior

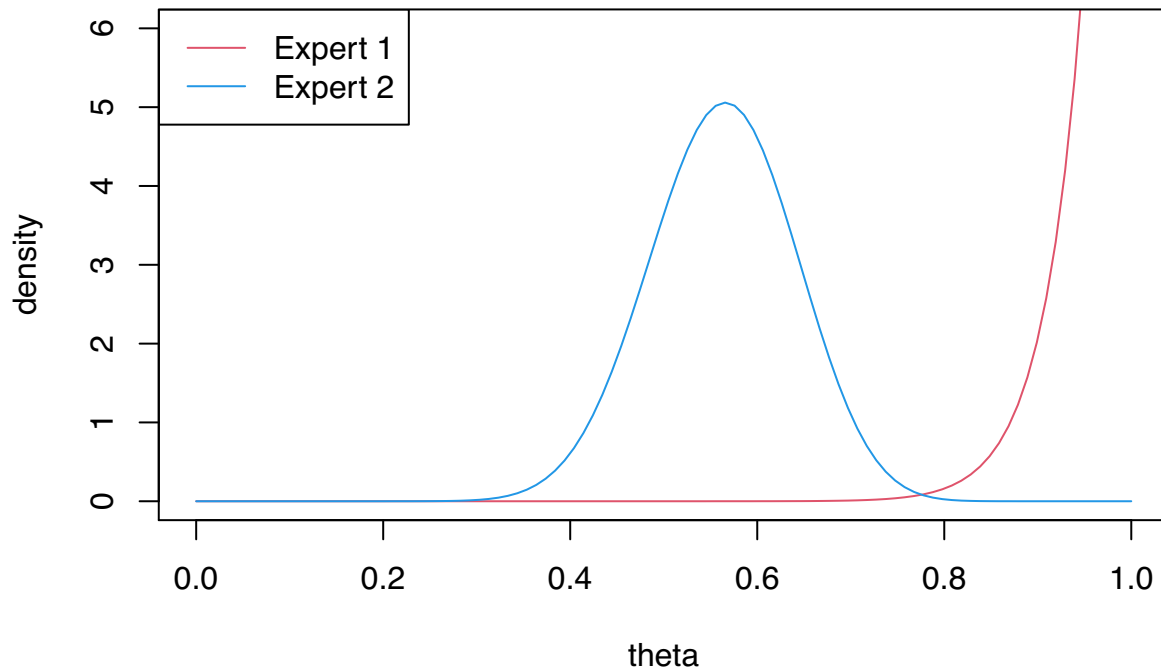
$$b = \frac{35}{2}, a = \frac{35}{2}$$

Expert 2's prior is uninformative since solving for the desired mean and standard deviation yields $a = b = 17.5$.

1.b)

The posterior for Expert 1 is relatively unchanged given all 5 mice test positive, since it expected 0.95 probability for each mouse, whereas the posterior for Expert 2 has a greater mean and lower variance.

Posteriors under each expert's prior



2.a)

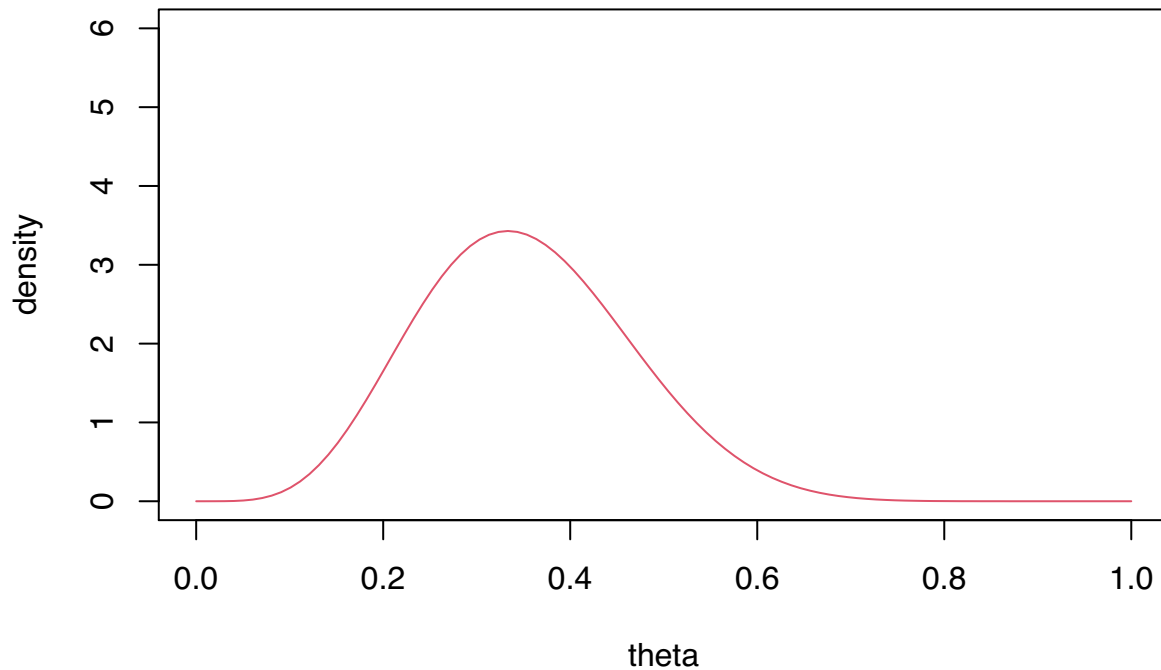
The Negative Binomial distribution is used to count the number of failures from independent trials with probability of success θ before recording the m th success.

2.b) $Y|\theta \sim \text{Neg Binom}(\theta, m)$, $\theta \sim \text{Beta}(a, b)$

$$\begin{aligned}
 p(\theta|Y) &\propto \theta^{a-1} (1-\theta)^{b-1} \binom{Y+m-1}{Y} \theta^m (1-\theta)^Y \\
 &\propto \theta^{a+m-1} (1-\theta)^{b+Y-1} \\
 &= \text{Beta}(a+m, b+Y)
 \end{aligned}$$

2.c)

Posterior when $m = 5$, $Y = 10$, $a = b = 1$



95% credible interval = (0.1519837, 0.5866206).

3)

$$p(\lambda|Y) = \frac{p(Y|\lambda)p(\lambda)}{\int p(Y|\lambda)p(\lambda)d\lambda}$$

$$\text{Top: } \frac{\lambda^Y e^{-\lambda}}{Y!} \cdot \frac{b^a}{\Gamma(a)} \lambda^{a-1} e^{-b\lambda} = \frac{b^a}{Y! \Gamma(a)} \lambda^{a+Y-1} e^{-(b+1)\lambda}$$

$$\text{Bottom: } \int \frac{\lambda^Y e^{-\lambda}}{Y!} \cdot \frac{b^a}{\Gamma(a)} \lambda^{a-1} e^{-b\lambda} d\lambda = \frac{b^a}{Y! \Gamma(a)} \frac{\Gamma(a+Y)}{(b+1)^{a+Y}} \int \frac{(b+1)^{a+Y}}{\Gamma(a+Y)} \lambda^{a+Y-1} e^{-(b+1)\lambda} d\lambda$$

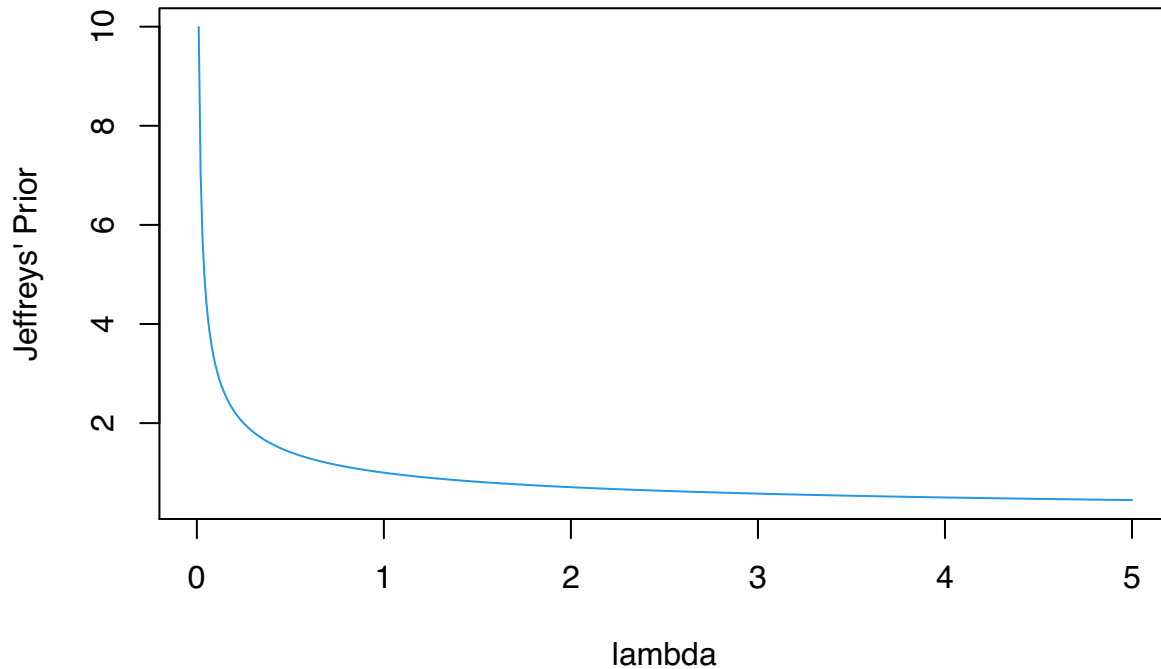
Gamma $\Rightarrow = 1$.

$$\text{Thus, } p(\lambda|Y) = \frac{\frac{b^a}{Y! \Gamma(a)} \lambda^{a+Y-1} e^{-(b+1)\lambda}}{\frac{b^a}{Y! \Gamma(a)} \frac{\Gamma(a+Y)}{(b+1)^{a+Y}}} = \frac{(b+1)^{a+Y}}{\Gamma(a+Y)} \lambda^{a+Y-1} e^{-(b+1)\lambda}$$

4.a) $\log(L(\lambda|Y)) \propto -\lambda - \ln(Y!) + \ln(\lambda)Y$

$\frac{\partial^2 \log(L(\lambda|Y))}{\partial \lambda^2} = -\frac{Y}{\lambda^2}$. Taking $-E(Y)$ gives $1/\lambda$. $\Rightarrow p(\lambda) \propto \lambda^{-1/2}$

Plot of Jeffreys' Prior for $\lambda \sim \text{Poisson}$



4.b)

No, this is not a proper prior since the integral does not converge. $\int_0^{\infty} \lambda^{-1/2} d\lambda$ doesn't converge, i.e. it's improper.

4.c)

$$p(\lambda|Y) = \frac{\lambda^Y e^{-\lambda}}{Y!} \lambda^{-1/2} = \frac{\lambda^{Y-1/2} e^{-\lambda}}{Y!} = \text{Gamma}(Y+1/2, 1)$$

$Y \geq 0$ ensures this is a proper posterior.