

1. a) $p(\theta|y) \propto p(y|\theta)p(\theta) \propto \theta^{2a} e^{-y\theta^2} p(\theta)$
 A class of conjugate prior is $p(\theta) = c(\alpha, \beta) \theta^\alpha e^{-\beta\theta^2}$ which is the Galenshore distribution.

b) $p(\theta) = \frac{2}{\Gamma(d)} \beta^d \theta^{2d-1} e^{-\beta\theta^2}$
 $p(\theta|y_1, \dots, y_n) \propto \prod_{i=1}^n p(y_i|\theta)p(\theta) \propto \theta^{2na} e^{-\theta^2 \sum y_i^2} \cdot \theta^{2d-1} e^{-\beta\theta^2} = \theta^{2na+2d-1} e^{-(\beta^2 + \sum y_i^2)\theta^2}$
 which is Galenshore $(na+d, \sqrt{\beta^2 + \sum y_i^2})$

c) $\frac{p(\theta_a|y_1, \dots, y_n)}{p(\theta_b|y_1, \dots, y_n)} = \left(\frac{\theta_a}{\theta_b}\right)^{2na+2d-1} e^{-(\beta^2 + \sum y_i^2)(\theta_a^2 - \theta_b^2)}$

$\sum y_i^2$ is a sufficient statistic.

d) $E[\theta|y_1, \dots, y_n] = \frac{\Gamma(na+d+\frac{1}{2})}{\sqrt{\beta^2 + \sum y_i^2} \Gamma(na+d)}$

e) $p(\tilde{y}|y_1, \dots, y_n) = \int p(\tilde{y}|\theta)p(\theta|y_1, \dots, y_n)d\theta \propto \int \theta^{2a} \tilde{y}^{2a-1} e^{-\theta^2 \tilde{y}^2} \cdot \theta^{2na+2d-1} e^{-(\beta^2 + \sum y_i^2)\theta^2} d\theta$
 $= \tilde{y}^{2a-1} \int \theta^{2(n+1)a+2d-1} e^{-(\beta^2 + \sum y_i^2 + \tilde{y}^2)\theta^2} d\theta$
 $\propto \tilde{y}^{2a-1} (\beta^2 + \sum y_i^2 + \tilde{y}^2)^{-(n+1)a-d}$

The integral can be derived from the density function that $\int y^{2a-1} e^{-\theta^2 y^2} dy = \frac{\Gamma(a)}{2\theta^{2a}}$

2. a) $\theta_A | x_A \sim \text{Gamma}(237, 20)$ Posterior mean is $\frac{237}{20}$, variance $\frac{237}{400}$
 $\theta_B | x_B \sim \text{Gamma}(118, 13)$ Posterior mean is $\frac{118}{13}$, variance $\frac{118}{169}$
 $P(\theta_A > \theta_B | x_A, x_B) = F_{\theta_A + \theta_B}(118, 237) \left(\frac{13}{20+13}\right) \approx 0.992$

3. $P(X=1) = \int p(X=1)p\pi(p)dp = \frac{a}{a+b}$
 so the variance of prior predictive distribution is $E[X^2] - (E[X])^2 = \frac{ab}{(a+b)^2}$

$p(X=1 | x_1, \dots, x_n) = \int p(X=1)p\pi(p|x_1, \dots, x_n)dp = \frac{a + \sum x_i}{a+b+n}$
 so the variance of posterior predictive distribution is $\frac{(a + \sum x_i)(b+n - \sum x_i)}{(a+b+n)^2}$

Note that the first variance depends on the ratio of $\frac{a}{b}$ while the other is $\frac{a + \sum x_i}{b+n - \sum x_i}$

4. a) $\int_0^1 \sqrt{1-x^2} dx = \int_0^{\frac{\pi}{2}} \cos y \sin y dy = \int_0^{\frac{\pi}{2}} \frac{1}{2} \sin 2y dy = \int_0^{\frac{\pi}{2}} \cos 2y dy = \pi$