Problem 1a:

Bayes' theorem gives for $y = (y_1, ..., y_n)$ that $P(\Theta|y) \propto P(y|\Theta) P(\Theta) \propto \Theta^{2n} \cdot \exp[-(\frac{1}{2} + \sum_{i=1}^{n} y_i)\Theta]$, which can be identified as the form of a gamma distribution with parameters $\alpha = 2n + 1$ and $\beta = \frac{1}{2} + \sum_{i=1}^{n} y_i$

Bayes' theorem gives for
$$x = (x_{11}, x_{n})$$
 that $P(\Theta|X) \propto P(X|\Theta) P(\Theta) \propto \exp\left[-n\Theta^{2} + 2\Theta \sum_{i=1}^{n} \ln x_{i}\right]$ If $\Theta|X \sim N(\mu_{n}, \sigma_{n}^{2})$, then $P(\Theta|X) \propto \exp\left[-\frac{1}{2\sigma_{n}^{2}} \cdot \Theta^{2} + \frac{\mu_{n}}{\sigma_{n}^{2}} \cdot \Theta\right]$.

Hence, the posterior distribution is
$$N(\mu_n, \sigma_n^2)$$
, where it is identified that $\mu_n = \frac{\sum_{i=1}^n \ln x_i}{n}$ and $\sigma_n^2 = \frac{1}{2n}$

Problem 36:

The Bayes point estimate is the posterior median given by
$$\mu_n$$
, i.e. $\hat{\Theta} = \mu_n = \frac{2}{5} = 0.4$

Problem 3c:

The marginal likelihood for
$$x = (x_1, ..., x_n)$$
 and $p(\theta) = \frac{1}{c}$ on an internal with length c is given by
$$p(x) = \int_{-\infty}^{\infty} p(x|\theta) p(\theta) d\theta = \frac{1}{c} \cdot \pi^{-n/2} \cdot \lim_{i=1}^{n} \left(\frac{1}{|x_i|}\right) \cdot \exp\left[-\frac{y_i}{i=1}(\ln x_i)^2\right]$$

$$= \exp\left[\frac{\mu_n^2}{2\sigma_n^2}\right] \cdot \int_{-\infty}^{\infty} \exp\left[-\frac{1}{2\sigma_n^2} \cdot (\theta - \mu_n)^2\right] d\theta$$

$$= \sigma_n \cdot \sqrt{2\pi}$$