

### Problem 1a:

Bayes' theorem gives for  $x$  that

$P(\theta|x) \propto p(x|\theta) p(\theta) \propto \theta^{2 + \sum_{i=1}^n x_i} \cdot \exp[-(0.5+n)\theta]$ , which can be identified as the form of a gamma distribution with parameters  $\alpha = 3 + \sum_{i=1}^n x_i$  and  $\beta = 0.5 + n$

### Problem 1b:

The Bayes point estimate is the posterior mean given by

$$\hat{\theta} = \frac{\alpha}{\beta} = \frac{78}{15.5} \approx 5.03$$

### Problem 1c:

The marginal likelihood for  $x$  is given by

$$P(x) = \int_0^{\infty} P(x|\theta) p(\theta) d\theta = \frac{1}{16 \cdot \prod_{i=1}^n x_i!} \cdot \underbrace{\int_0^{\infty} \theta^{\alpha-1} \exp[-\beta\theta] d\theta}_{= \frac{\Gamma(\alpha)}{\beta^{\alpha}}}$$



### Problem 3a:

From lecture 2, slide 4 we have that

$$\frac{1}{\tau_n^2} = \frac{n}{\sigma^2} + \frac{1}{\tau_0^2} \iff \frac{1}{2^2} = \frac{10}{50} + \frac{1}{\tau_0^2},$$

which gives  $\tau_0^2 = 20$  and  $w = \frac{\frac{10}{50}}{\frac{10}{50} + \frac{1}{20}} = 0.8$ , so that

$$\mu_n = 92 = 0.8 \cdot 90 + (1 - 0.8) \cdot \mu_0 \quad \text{gives} \quad \mu_0 = 100$$

Hence,  $\mu \sim N(\mu_0 = 100, \tau_0^2 = 20)$