

3. In a mass production process items are made to a nominal weight of 1.0 kg.

$$\begin{aligned}
 1 &= \int_0^{10} \frac{C}{100} dx + \int_{10}^{\infty} C x^{-2} dx \\
 &= \frac{C}{100} x \Big|_0^{10} + C \frac{x^{-2+1}}{-2+1} \Big|_{10}^{\infty} \\
 &= \frac{C}{100} \times 10 + \frac{C}{10} = \\
 &= C \left[\frac{1}{10} + \frac{1}{10} \right] \Rightarrow C = 5
 \end{aligned}$$

b) we observe that 10 items and

i) The density function for a random variable X with a uniform distn

$(0, \theta)$ is $P(X/\theta) = \frac{1}{\theta} \cdot 0 \leq x \leq \theta$

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ii) It follows that the likelihood is

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The range follows from the fact that

$$\theta \geq x_1, \theta \geq x_2 \text{ and so on}$$

Since θ is larger than each x_i .

It must be larger than the

largest of the x_i , for the data set

given above $\max x_i = 4.9$

$$P(n/\theta) = \begin{cases} \frac{1}{\theta^{10}} & \text{for } \theta \geq 4.9 \\ & \text{for } \theta < 4.9 \end{cases}$$

$$\Rightarrow P(\theta/n) \propto \begin{cases} \frac{1}{100} \times \frac{1}{\theta^{10}} & \text{for } 4.9 \leq \theta \leq 10 \\ 50^{-2} \times \frac{1}{\theta^{10}} & \text{for } \theta > 10 \end{cases}$$

III) The normalizing constant for the posterior density

$$1 = D \left[\int_{4.9}^{10} \frac{1}{20 \theta^{10}} d\theta + \int_{10}^{\infty} \frac{1}{\theta^{12}} d\theta \right]$$

$$= D = \frac{1}{3.4106 \times 10^{-9}}$$

4) a) The pmf function

$$J = 1 \Rightarrow P(n/\theta) = \theta n^{\theta-1}$$

The likelihood $\Rightarrow 0 < n < 1$

$$P(n/\theta) = \prod_{i=1}^n \theta n_i^{\theta-1} = \theta^n \prod_{i=1}^n n_i^{\theta-1}$$

The prior is $P(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\theta\beta}$

By Bayes theorem the posterior distⁿ is given by

$$P(\theta|n) \propto \theta^n \left(\prod_{i=1}^n x_i^{\theta-1} \right) \propto \theta^{n+\alpha-1} e^{-\theta\beta} \exp\left(\log \prod_{i=1}^n \theta x_i^{\theta-1}\right)$$

$$= \theta^{n+\alpha-1} e^{-\theta\beta} \exp\left(\theta \sum_{i=1}^n \log x_i\right)$$

$$= \theta^{n+\alpha-1} e^{-\theta\beta} \exp\left(\sum_{i=1}^n (\theta-1) \log x_i\right)$$

$$\propto \theta^{n+\alpha-1} e^{-\theta\beta} \exp\left(\theta \sum_{i=1}^n \log x_i\right)$$

$$= \theta^{n+\alpha-1} \exp\left\{-\theta \left(\beta - \sum_{i=1}^n \log x_i\right)\right\}$$

∴ kernel of Gamma $\left(n+\alpha, \frac{1}{\beta - \sum_{i=1}^n \log x_i}\right)$

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Since the prior and posterior distn belong to the same Gamma family of distn $\text{Gamma}(\alpha, 1/B)$ it is conjugate prior for θ .

The prior density function when

$$\theta = 2 \text{ is } P(n/\delta) = 2\delta^{2n} \quad 0 \leq n < \delta$$

\Rightarrow The likelihood for x_1, x_2, \dots, x_n is

$$P(n/\delta) = \prod_{i=1}^n P(x_i/\delta)$$

$$= \prod_{i=1}^n 2\delta^{2x_i} \propto \delta^{2n} \max x_i < \delta$$

equivalently $\min x_i^{-1} > \delta$

6) The posterior is

$$P(\delta/n) \propto \delta^{-1} \delta^{2n} \min x_i^{-1} > \delta$$

To find the normalizing constant let $\min x_i^{-1} = t$ then

$$1 = \int_0^t C \delta^{-2n+1} d\delta = C \frac{\delta^{-2n+2}}{-2n+2} \Big|_0^t = C \frac{t^{-2n+2}}{-2n+2} \Rightarrow C = \frac{2n-2}{t^{-2n+2}}$$

6. Data $n=12$ $\pi = 16.35528$

$$p(\mu/\pi) \propto (\sqrt{2\pi})^{-n} \exp\left[-\frac{1}{2} \sum_{i=1}^n (\mu_i - \mu)^2\right]$$

$$= (\sqrt{2\pi})^{-n} \exp\left[-\frac{1}{2} \sum_{i=1}^n (\mu_i^2 - 2\mu_i\mu + \mu^2)\right]$$

$$\propto \exp\left(-\frac{1}{2} n\mu^2 - 2\mu \sum_{i=1}^n \mu_i\right)$$

$$= \exp\left(-\frac{1}{2} \left\{ n(\mu^2 - 2\mu\bar{\pi}) \right\}\right)$$

$$\propto \exp\left[-\frac{n}{2} (\mu - \bar{\pi})^2\right] \text{ kernel of Normal } \left(\bar{\pi}, \frac{1}{n}\right)$$

\Rightarrow Posterior: Normal $(\bar{\pi}, \frac{1}{n}) = \text{Normal}$

$$90\% \text{ HPD for } \mu: 16.35528 \pm 1.65 \sqrt{\frac{1}{12}}$$

$$\approx (15.88)$$