

Given

$$D = \{x_1, x_2, \dots, x_n\} \sim \text{poisson}$$

1) Prior = exponential, with mean β

$$\text{and } \beta = X_{\text{mme}} = \lambda \quad \text{--- (1)}$$

$$\text{Exponential}(X) = \lambda e^{-\lambda x}$$

$$E[X] = 1/\lambda \quad \text{--- (2)}$$

from (1) and (2)

$$\lambda = 1/\beta$$

$$2) \text{ Likelihood} = \prod_{i=1}^n \frac{\lambda^{x_i} e^{-\lambda}}{x_i!} = \frac{e^{-n\lambda} \lambda^{\sum_i x_i}}{\prod_i x_i!}$$

3) Posterior \propto likelihood \times prior

$$\begin{aligned} & \frac{e^{-n\lambda} \lambda^{\sum_i x_i}}{\prod_{i=1}^n x_i!} \cdot \frac{e^{-\lambda/\beta}}{\beta} \\ &= e^{-\lambda(n+1/\beta)} \lambda^{\sum_i x_i} \quad \text{--- (3)} \end{aligned}$$

$$4) \text{ Gamma } (d, \beta) = \lambda^{d-1} e^{-\lambda/\beta} \quad - (4)$$

Comparing (3) and (4) we have

$$d-1 = \sum_{i=1}^n x_i \Rightarrow d = \sum_{i=1}^n x_i + 1$$

$$n + \frac{1}{\beta_P} = \frac{1}{\beta} \Rightarrow \beta = \frac{1}{\left(n + \frac{1}{\beta_P}\right)}$$

from posterior.