

$X_i \sim \text{Poisson}(\theta)$, i.i.d. because we have a random sample.

$$n = 74, \quad \sum_{i=1}^{74} x_i = 192$$

a) statistical model.

X is the r.v. that describes the number of problems that a student solves.

$$\begin{aligned} f(x_1, \dots, x_n | \theta) &= f_n(\underline{x} | \theta) = \prod_{i=1}^n f(x_i | \theta) = \prod_{i=1}^n \frac{e^{-\theta} \theta^{x_i}}{x_i!} \\ &= \frac{e^{-n\theta} \theta^{\sum x_i}}{\prod_{i=1}^n x_i!}, \quad \theta > 0, \quad x_i \in \{0, 1, 2, \dots\} \end{aligned}$$

$$\Omega = \mathbb{R}^+$$

b) a statistical inference problem is finding the estimate of the mean.

c) find the MLE of interest: mean and coef. of variation of the # of problems that students solve.

$$\begin{aligned} E(X) &= \theta \\ CV(X) &= \frac{\sqrt{\text{Var}(X)}}{|E(X)|} = \frac{\sqrt{\theta}}{\theta} = \frac{1}{\sqrt{\theta}} \end{aligned}$$

i) Let's find the MLE of θ .

$$f_n(\underline{x} | \theta) = \frac{e^{-n\theta} \theta^{\sum x_i}}{\prod_{i=1}^n x_i!}$$

$$\begin{aligned} L(\theta) &= \log f_n(\underline{x} | \theta) \\ &= \log \left(\frac{e^{-n\theta} \theta^{\sum x_i}}{\prod_{i=1}^n x_i!} \right) \end{aligned}$$

(log = ln)

$$= -n\theta + \sum_{i=1}^n x_i \log(\theta) - \log \prod_{i=1}^n x_i!$$

$$\frac{d}{d\theta} L(\theta) = -n + \frac{\sum_{i=1}^n x_i}{\theta} = 0$$

$$\Rightarrow \frac{\sum_{i=1}^n x_i}{\theta} = n$$

$$\Rightarrow \theta = \frac{\sum x_i}{n}$$

$$\frac{d}{d\theta} \left[-n + \frac{\sum x_i}{\theta} \right] = -\frac{\sum x_i}{\theta^2}$$

$$\frac{d^2}{d\theta^2} L(\theta) = -\frac{\sum x_i}{\theta^2} < 0 \quad \forall x_i, \theta \quad \left| \begin{array}{l} \text{checking that} \\ \theta = \frac{\sum x_i}{n} \text{ is a maximum} \end{array} \right.$$

Therefore $\hat{\theta} = \frac{\sum X_i}{n}$ is the M.L. estimator

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$$\hat{\theta} = \frac{\sum x_i}{n} = \frac{192}{74}$$

Answer: the MLE of the mean of the # of problems that students solve is $\frac{192}{74} = 2.59$

Then the MLE for $CV(X)$ is $\frac{1}{\sqrt{\frac{\sum X_i}{n}}}$ (estimator)

the maximum likelihood estimate of $CV(X)$ is

$$\sqrt{\frac{1}{\sum x_i / n}} = \sqrt{\frac{1}{192/74}} = \frac{1}{\sqrt{2.59}}$$

↓) θ is r.v. and ~~$\theta \sim \text{Gamma}$~~ θ has a gamma distribution with mean and variance equal to 2.

$$\theta \sim \text{Gamma}(a, b)$$

$$E(\theta) = \frac{a}{b} = 2$$

$$\text{Var}(\theta) = \frac{a}{b^2} = 2$$

$$\frac{1}{b} \cdot 2 = 2 \Rightarrow \boxed{\begin{matrix} b=1 \\ a=2 \end{matrix}}$$

So $\theta \sim \text{Gamma}(2, 1)$

Under square error loss function

Bayes estimate of $E(X)$ is $E(\theta | x)$

Bayes estimate of $\text{CV}(X)$ is $E\left(\frac{1}{\theta} | x\right)$

• Let's find the posterior distr. of θ given data.

$$\begin{aligned} \pi(\theta | x) &\propto f_n(x | \theta) \xi(\theta) \\ &= \frac{e^{-n\theta} \theta^{\sum x_i}}{\prod_{i=1}^n x_i!} \cdot \frac{1}{\Gamma(2)} \theta^{2-1} e^{-1 \cdot \theta} \end{aligned}$$

likelihood Gamma distr.

$$\propto \theta^{2 + \sum x_i - 1} e^{-(n+1)\theta}$$

looks like $\text{Gamma}\left(\frac{2 + \sum x_i}{194}, \frac{n+1}{75}\right)$

So,

Bayes estimate for $E(X)$ is $\frac{2 + \sum x_i}{n+1} = \frac{2 + 192}{74+1} = \frac{194}{75}$

Bayes estimate for $CV(x)$ is

$$E\left(\frac{1}{\sqrt{\theta}} \mid x\right) = \int_0^{\infty} \frac{1}{\sqrt{\theta}} \varphi(\theta \mid x) d\theta$$

$$= \int_0^{\infty} \theta^{-1/2} \frac{(75)^{194}}{\Gamma(194)} \theta^{194-1} e^{-75\theta} d\theta$$

$$= \frac{(75)^{194}}{\Gamma(194)} \int_0^{\infty} \theta^{194-1/2-1} e^{-75\theta} d\theta$$

looks like Gamma $(194-1/2, 75)$

$$= \frac{(75)^{1/2} \Gamma(193.5)}{\Gamma(194)}$$

$$\rightarrow = \frac{(75)^{194}}{\Gamma(194)} \int_0^{\infty} \frac{\Gamma(194-1/2)}{(75)^{194-1/2} \Gamma(194-1/2)} \theta^{194-1/2-1} e^{-75\theta} d\theta$$

integrates to 1

$$= \frac{(75)^{194}}{\Gamma(194)} \frac{\Gamma(194-1/2)}{(75)^{194-1/2}}$$

$$= \frac{75^{1/2}}{\Gamma(194)} \Gamma(194-1/2)$$

e) distr. of a new observation given $X_1 = x_1$. $X_n = x_n$

$$f(x_{n+1} | x_1, x_n) = \int_0^{\infty} f(x_{n+1} | \theta) \xi(\theta | x) d\theta \quad \text{Gamma}(193.5, 75)$$

$$= \int_0^{\infty} e^{-\theta} \frac{\theta^{x_{n+1}}}{x_{n+1}!} \frac{(75)^{193.5}}{\Gamma(193.5)} \theta^{193.5-1} e^{-75\theta} d\theta$$

$$= \frac{(75)^{193.5}}{\Gamma(193.5)} \frac{1}{x_{n+1}!} \int_0^{\infty} \theta^{193.5-1} e^{-76\theta} d\theta$$

$$= \frac{(75)^{193.5}}{\Gamma(193.5)} \frac{1}{x_{n+1}!} \frac{\Gamma(193.5)}{(76)^{193.5}} \underbrace{\int_0^{\infty} \frac{(76)^{193.5}}{\Gamma(193.5)} \theta^{193.5-1} e^{-76\theta} d\theta}_1$$

$$= \frac{(75)^{193.5}}{(76)^{193.5}} \frac{1}{x_{n+1}!}$$