

Gracim Bon-Flavim

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1.a) The probability is modelled by a Poisson distribution with $\lambda = 10$ (10 messages per hour). knowing that we need to compute

$$\frac{10^5}{5!} \cdot e^{-10} + \frac{10^6}{6!} \cdot e^{-10} + \dots = 1 - \text{poisscdf}(4, 10) = 0.97075$$

b) $P(x \geq 8 | x \geq 6)$?

$$P(x \geq 8 | x \geq 6) = \frac{P(x \geq 8 \cap x \geq 6)}{P(x \geq 6)} = \frac{P(x \geq 8)}{P(x \geq 6)}$$
$$= \frac{1 - \text{poisscdf}(7, 10)}{1 - \text{poisscdf}(5, 10)} = \frac{0.77928}{0.93291} = 0.83585$$

c) For a ~~Poisson~~ variable with Poisson distribution the expected value is equal to its parameter λ which models the average number of events in a time interval.

If 1 message arrives in 6 minutes then 15 messages should arrive in 90 minutes.

$$E(x) = 15 \text{ where } x \sim \text{pois with } \lambda = 15$$

Q. $X \in N(0, 1)$

$$Y = 2X - 7$$

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$g(x) = 2x - 7$ is a strictly increasing function and also differentiable.

$$g'(x) = 2$$

$$g^{-1}(x) = \frac{x+7}{2}$$

$$g(\mathbb{R}) = \mathbb{R}$$

$$\text{Ansatz } f_x = \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{x^2}{2}}$$

$$f_y = \frac{\frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{(x+7)^2}{2}}}{2} = \frac{1}{\sqrt{8\pi}} \cdot e^{-\frac{(x+7)^2}{8}}$$

$$\text{By applying } f_y = \begin{cases} f_x(g^{-1}(y)) \\ \frac{1}{|g'(g^{-1}(y))|} \end{cases}, y \in g(\mathbb{R})$$

$$\therefore Y \notin g(\mathbb{R})$$

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3. a) If we solve $\sigma_1 = \bar{\sigma}_1$

where $\sigma_1 = E(x)$ and $\bar{\sigma}_1$ is the mean of all samples.

$$\text{Zp} \quad E(x) = \bar{x}$$

$$3\mu = \bar{x}$$

$\hat{\mu} = \frac{\bar{x}}{3}$ — estimator from method of moments

b) Only one unknown \Rightarrow only one equation

$$\frac{\partial \ln L(x_1, \dots, x_n, \mu)}{\partial \mu} = 0$$

$$\begin{aligned} L(x_1, x_2, \dots, x_n, \mu) &= \prod_{i=1}^n f(x_i; \mu) \\ &= \prod_{i=1}^n \frac{1}{3\mu} \cdot e^{-\frac{x_i}{3\mu}} \\ &= \frac{n}{(3\mu)^n} \cdot e^{-\sum_{i=1}^n \frac{x_i}{3\mu}} \end{aligned}$$

$$\begin{aligned} \ln L &= -n \cdot \ln 3\mu + \sum_{i=1}^n \frac{-x_i}{3\mu} \cdot \ln 3\mu \\ &= -n \cdot \ln 3\mu - \frac{3}{3\mu} \cdot \sum_{i=1}^n x_i \end{aligned}$$

$$\frac{\partial \ln L}{\partial \mu} = -n \cdot \frac{3}{\mu} + \frac{3}{3\mu^2} \cdot \sum_{i=1}^n x_i$$

$$\frac{\partial L}{\partial \mu} = 0 \Rightarrow -\frac{n}{\mu} + \frac{3}{3\mu^2} \cdot \bar{x} = 0 \quad | \cdot \mu^2$$

$$-\frac{n}{\mu} + \frac{1}{\mu} \bar{x} = 0$$

$$\bar{\mu} = \frac{\bar{x}}{3n} \text{ - estimator for max likelihood}$$

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c) $\bar{\mu}$ is absolutely correct if

$$E(\bar{\mu}) = \mu \quad \text{and} \quad \lim_{n \rightarrow \infty} V(\bar{\mu}) = 0$$

$$E(\bar{\mu}) = E\left(\frac{\bar{x}}{3n}\right) = \frac{1}{3n} E(\bar{x}) = \frac{1}{3n} \cdot E(x) = \frac{1}{3n} \cdot 3\mu = \frac{\mu}{n}$$

$$V(\bar{\mu}) = V\left(\frac{\bar{x}}{3n}\right) = \left(\frac{1}{3n}\right)^2 \cdot V(\bar{x}) = \frac{1}{9n} \cdot V(x) \\ = \frac{1}{9n} \cdot \frac{8\mu^2}{n} \\ = \frac{\mu^2}{n^2} \rightarrow 0$$

Only one requirement is met $\Rightarrow \bar{\mu}$ is not an absolutely correct estimator



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d) $f(x, \mu) = \frac{1}{3\mu} e^{-\frac{x}{3\mu}}$

We will use the Neyman-Pearson Lemma

$H_0: \mu = 1$

$H_A: \mu = 2$

$$RL = \left\{ \frac{L(\mu_1)}{L(\mu_0)} \geq k_a \right\}$$

$$\begin{aligned} \frac{L(\mu_1)}{L(\mu_0)} &= \frac{(3\mu_1)^{-n} \cdot e^{-\sum_{i=1}^n \frac{x_i}{3\mu_1}}}{(3\mu_0)^{-n} \cdot e^{-\sum_{i=1}^n \frac{x_i}{3\mu_0}}} \\ &= \left(\frac{\mu_1}{\mu_0} \right)^n \cdot e^{-n \sum_{i=1}^n \frac{x_i}{3\mu_0} + \sum_{i=1}^n \frac{x_i}{3\mu_1}} \\ &= \left(\frac{2}{1} \right)^n \cdot e^{-n \sum_{i=1}^n \frac{\mu_1 - \mu_0}{3\mu_0 \mu_1} \cdot x_i} \\ &= \cancel{e} \cdot \frac{2^n}{2} \cdot n! \cdot \sum_{i=1}^{n-1} x_i \\ &= 2^n \cdot n! \end{aligned}$$

$$RL = \left\{ 2^n \cdot n! \geq k_a \right\} = \left\{ \bar{x}_n \geq k_a \right\}$$

$$k_a = 6 \ln(k_a \cdot 2^n)$$

$$\alpha = P(\bar{X}_n \in \text{IR} | H_0) = P(\bar{X} \geq k_\alpha | \mu=1) \quad \text{Gracum bon-Theorem}$$

* By C.L.T. $\frac{\bar{X} - 3\mu}{\sqrt{n+3}\mu} \rightarrow N(0,1)$

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$$P(\bar{X} \geq k_\alpha | \mu=1) = P\left(\frac{\bar{X} - 3\mu_0}{\sqrt{n+3}\mu_0}, \frac{k_\alpha - 3\mu_0}{\sqrt{n+3}\mu_0} | \mu=1\right)$$

$$= P(Z \geq \frac{k_\alpha - 3\mu_0}{\sqrt{n+3}\mu_0} | \mu=1)$$

$$= P(Z \geq \frac{k_\alpha - 3}{3\sqrt{n}})$$

$$= \cancel{P(Z \geq)} \int_{\frac{k_\alpha - 3}{3\sqrt{n}}}^{\infty} N(0,1) dx$$

$$= \underbrace{\int_{6\ln(k_\alpha \cdot 2^n) - 3}^{\infty} N(0,1) dx}_{8\sqrt{n}}$$

Now write k_α as a function of α and you have a rejection region for a most powerful test