

Computational Statistics Project 2

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For the Maximum Likelihood, from the pdf I will construct the Likelihood function. Then the log Likelihood and differentiate with respect to θ .

Set the derivative to zero to find θ .

Then I ~~will~~ use Newton-Raphson method to calculate $\hat{\theta}$ numerically.

$$\text{pdf} \quad f(x; \theta) = \frac{\theta^3}{\theta+1} \times (1 + \frac{x}{2}) \exp(-\theta x)$$

$$\text{Likelihood} \quad L(\theta) = L(\theta|x) = \prod_{i=1}^n f(x_i; \theta) = \prod_{i=1}^n \left(\frac{\theta^3}{\theta+1} \times (1 + \frac{x_i}{2}) \exp(-\theta x_i) \right) =$$

$$= \left(\frac{\theta^3}{\theta+1} \right)^n \prod_{i=1}^n x_i \cdot \prod_{i=1}^n (1 + \frac{x_i}{2}) \exp(-\theta \sum_{i=1}^n x_i)$$

$$\text{log Likelihood} \quad l(\theta) = \log(L(\theta)) = \log \left[\left(\frac{\theta^3}{\theta+1} \right)^n \cdot \prod_{i=1}^n x_i \cdot \prod_{i=1}^n (1 + \frac{x_i}{2}) \exp(-\theta \sum_{i=1}^n x_i) \right]$$

$$= \log \left(\frac{\theta^3}{\theta+1} \right)^n + \log \prod_{i=1}^n x_i + \log \prod_{i=1}^n (1 + \frac{x_i}{2}) + \log (\exp(-\theta \sum_{i=1}^n x_i)) =$$

$$= n(\log \theta^3 - \log(\theta+1)) + \sum_{i=1}^n \log x_i + \sum_{i=1}^n \log(1 + \frac{x_i}{2}) - \theta \sum_{i=1}^n x_i =$$

$$= n(3 \log \theta - \log(\theta+1)) + \sum (\log x_i + \log(1 + \frac{x_i}{2})) - \theta \sum_{i=1}^n x_i$$

take derivatives

$$\frac{\partial l(\theta)}{\partial \theta} = 0 \Leftrightarrow n \left(\frac{3}{\theta} - \frac{1}{1+\theta} \right) - \sum_{i=1}^n x_i = 0$$

$$\frac{3}{\theta} - \frac{1}{1+\theta} = \frac{\sum_{i=1}^n x_i}{n} = \bar{x} \Leftrightarrow \bar{x} = \frac{3(\theta+1) - \theta}{\theta(1+\theta)}$$

$$\Leftrightarrow \bar{x} = \frac{2\theta+3}{\theta^2+\theta} \Leftrightarrow \bar{x}(\theta^2+\theta) = 2\theta+3$$

$$\Leftrightarrow \bar{x}\theta^2 + \bar{x}\theta - 2\theta - 3 = 0 \Leftrightarrow \bar{x}\theta^2 + (\bar{x}-2)\theta - 3 = 0$$

$$\frac{\partial^2 l(\theta)}{\partial \theta^2} = 0 \Leftrightarrow 2 \cdot \bar{x}\theta + (\bar{x}-2) = 0$$

Because this is challenging to solve analytically, I will use N-R method as I mentioned before: I set $F(\theta) = \bar{x}\theta^2 + (\bar{x}-2)\theta - 3$ and set it to zero, to solve it numerically.

1) I initialize $\theta^{(0)}$, I set it to be 1

2) $\theta^{(r+1)} = \theta^{(r)} - \frac{F(\theta^{(r)})}{F'(\theta^{(r)})}$ 3) Repeat while $|\theta^{(r+1)} - \theta^{(r)}| < \epsilon$

After running the algorithm for various initial values I found the $\theta = 0.797472$ the only acceptable value for MLE.

Then for the standard error estimation using parametric ~~boot~~ bootstrap, i will use rejection method to produce a sample from the original pdf. my envelope function will be $\text{Gamma}(2, 0.6)$ and $M = 1.1$ both of which I found graphically in R trying out different values.

- 1) ~~take~~ I take random sample from $g \sim \text{Gamma}(2, 0.6)$
- 2) I take $y \sim \text{unif}(0, M * g(x))$
- 3) if $y < f(x, \theta)$ keep x .
~~else~~ else, go back to step 1.

In the bootstrap i produce a new sample everytime and calculate and keep the theta in a vector. ~~And~~

finally i calculate the standard error from the standard deviation of the bootstrap thetas vector.

$$S.E. = 0.063$$

To examine if the data come from the given distribution we need a test statistic for goodness of fit, for each simulated sample.

H_0 : the data come from the specific distribution.

H_1 : not H_0 .

i compare the theoretical and empirical Cdf.

The estimated p-value is $0.998 > 0.05$
so we do not reject the null hypothesis.
The data could plausibly ~~have~~ have come
from the specified distribution.

to find the CDF:

$$f(x, \theta) = \frac{\theta^3}{\theta+1} x \left(1 + \frac{x}{2}\right) e^{-\theta x}$$

$$F_X(x, \theta) = \int_0^x f(t, \theta) dt = \int_0^x \frac{\theta^3}{\theta+1} t e^{-\theta t} dt + \int_0^x \frac{\theta^3}{\theta+1} \frac{t^2}{2} e^{-\theta t} dt =$$

$$= \frac{\theta^3}{\theta+1} \left(\int_0^x t e^{-\theta t} dt + \int_0^x \frac{t^2}{2} e^{-\theta t} dt \right) = (*)$$

$$= \frac{\theta^3}{\theta+1} (I_1 + I_2)$$

$$I_1 = \int_0^x t e^{-\theta t} dt = \frac{x \cdot e^{-\theta x}}{-\theta} - \int_0^x \frac{e^{-\theta t}}{-\theta} dt = -\frac{x \cdot e^{-\theta x}}{\theta} + \frac{1}{\theta} \int_0^x e^{-\theta t} dt =$$

$$= -\frac{x e^{-\theta x}}{\theta} - \frac{1}{\theta^2} e^{-\theta t} \Big|_0^x = -\frac{x e^{-\theta x}}{\theta} - \frac{1}{\theta^2} e^{-\theta x} + \frac{1}{\theta^2}$$

$$I_2 = \int_0^x \frac{t^2}{2} e^{-\theta t} dt = \frac{1}{2} \int_0^x t^2 \cdot e^{-\theta t} dt = \textcircled{1}$$

$$= \frac{1}{2} \left(\frac{t^2}{2} \cdot \frac{e^{-\theta t}}{-\theta} - \int_0^x \frac{e^{-\theta t}}{-\theta} \cdot 2t dt \right) = \frac{1}{2} \left(-\frac{t^2}{\theta} e^{-\theta t} + \frac{2}{\theta} \int_0^x e^{-\theta t} dt \right) =$$

$$= \frac{1}{2} \left[-\frac{x^2}{\theta} e^{-\theta x} + \frac{2}{\theta} \left(-\frac{x e^{-\theta x}}{\theta} - \frac{1}{\theta^2} e^{-\theta x} + \frac{1}{\theta^2} \right) \right] =$$

$$= \frac{1}{2} \left(-\frac{x^2}{\theta} e^{-\theta x} - \frac{2x}{\theta^2} e^{-\theta x} - \frac{2}{\theta^3} e^{-\theta x} + \frac{2}{\theta^3} \right) =$$

$$= -\frac{x^2}{2\theta} \cdot e^{-\theta x} - \frac{x}{\theta} e^{-\theta x} - \frac{1}{\theta^3} e^{-\theta x} + \frac{1}{\theta^3} \quad \textcircled{2}$$

$$\text{So } \textcircled{1}, \textcircled{2} \quad \textcircled{*} = \frac{\theta^3}{\theta+1} \left(-\frac{x}{\theta} e^{-\theta x} - \frac{1}{\theta^2} e^{-\theta x} + \frac{1}{\theta^2} - \frac{x^2}{2\theta} e^{-\theta x} - \frac{x}{\theta} e^{-\theta x} - \frac{1}{\theta^3} e^{-\theta x} + \frac{1}{\theta^3} \right)$$

So this is our CDF.