



Maximum Likelihood Estimation

Defn: The likelihood function is

$$L(\theta) = \prod_{j=1}^n \text{pr}(X_j \in A_j | \theta)$$

where  $A_j$  results from observing the random variable  $X_j$ . If  $A_j$  is a single point and the distribution is continuous, then  $\text{pr}(X_j \in A_j | \theta)$  is interpreted as the probability density  $f_n$  evaluated at that point.

Ex (1) p. 231

Suppose that the data in Data Set B, introduced in ch. 10 and reproduced here as Table 11.1, were such that the exact value of all observations above 250 was unknown. All that is known is that the value was greater than 250. Determine the maximum likelihood estimate of  $\theta$  for an exponential distribution.

Ans:

Table 11.1 Data Set B

27	82	115	126	155	161	243	294	340	384
457	680	855	877	974	1,193	1,340	1,884	2,558	15,743

For the first seven values, the set  $A_j$  contains the single point equal to the observation  $x_j$ .

$$\begin{aligned} \Rightarrow f(27)f(82)\dots f(243) \\ = \theta^{-1}e^{-27/\theta} \theta^{-1}e^{-82/\theta} \dots \theta^{-1}e^{-243/\theta} \\ = \theta^{-7}e^{-909/\theta} \end{aligned}$$

For each of the final 13 terms, the set  $A_j$  is the interval from 250 to infinity and, therefore,  $\text{pr}(X_j \in A_j) = \text{pr}(X_j > 250) = e^{-250/\theta}$

$$\begin{aligned} X \sim \text{exp}(\theta) \\ f(x) = \frac{e^{-x/\theta}}{\theta} \\ S(x) = e^{-x/\theta} \end{aligned}$$

$$\therefore L(\theta) = \theta^{-7} e^{-909/\theta} (e^{-250/\theta})^{13} = \theta^{-7} e^{-4859/\theta}$$

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 $l(\theta) = -7 \ln \theta - 4159 \theta^{-1}$   
 which is known as log likelihood fn

$$\Rightarrow l'(\theta) = -7\theta^{-1} + 4159\theta^{-2} = 0$$

$$\Rightarrow \hat{\theta} = \frac{4159}{7} = 594.14$$

which is the likelihood estimate of the parameter  $\theta$ .

Note: clearly,  $l''(\hat{\theta})$  is -ve quantity, i.e.  $\hat{\theta}$  is a maximum estimate for the parameter  $\theta$ .

### Individual Data

Consider the special case where the value of each observation is recorded, it is easy to write the likelihood fn:

$$L(\theta) = \prod_{j=1}^n f_{x_j}(x_j|\theta) \quad (\text{This for Complete data})$$

and log likelihood fn:

$$l(\theta) = \sum_{j=1}^n \ln f_{x_j}(x_j|\theta)$$

Ex. 2 p. 232

Using Data set B, determine the maximum likelihood estimator for an exponential distribution, for a gamma dist<sup>n</sup> where  $\alpha$  is known to equal 2, and for a gamma dist<sup>n</sup> where both parameters are unknown.

Ans:

\* ① For exp dist<sup>n</sup>:

$$L(\theta) = \prod_{j=1}^n \frac{e^{-x_j/\theta}}{\theta}$$

$$\therefore l(\theta) = \sum_{j=1}^n (-\ln \theta - x_j \theta^{-1})$$

$$\therefore l(\theta) = -n \ln \theta - n \bar{x} \theta^{-1}, \quad \bar{x} = \frac{\sum_{j=1}^n x_j}{n}$$

$$\text{Let } l'(\theta) = 0 \Rightarrow l'(\theta) = -n\theta^{-1} + n\bar{x}\theta^{-2} = 0$$

$$\Rightarrow n\theta = n\bar{x} \Rightarrow \therefore \hat{\theta} = \bar{x}$$

From Data Set B,  $\hat{\theta} = 1424.4$  and  $l(\hat{\theta}) = -165.23$

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\* ② For the gamma dist<sup>n</sup> with  $\alpha=2$ :

$$f(x|\theta) = \frac{x^{\alpha-1} e^{-x/\theta}}{\Gamma(\alpha) \theta^\alpha}$$

$$f(x|\theta) = x \theta^{-2} e^{-x/\theta}$$

$$\ln f(x|\theta) = \ln x - 2 \ln \theta - x \theta^{-1}$$

$$\therefore l(\theta) = \sum_{j=1}^n \ln x_j - 2n \ln \theta - \theta^{-1} \sum_{j=1}^n x_j$$

$$\therefore l(\theta) = \sum_{j=1}^n \ln x_j - 2n \ln \theta - n \bar{x} \theta^{-1}, \quad \sum_{j=1}^n x_j = n \bar{x}$$

$$\Rightarrow l'(\theta) = -2n \theta^{-1} + n \bar{x} \theta^{-2} = 0$$

$$\Rightarrow 2 \theta^{-1} = \bar{x} \theta^{-2}$$

$$\therefore \hat{\theta} = \frac{\bar{x}}{2} \quad (\text{i.e. } \hat{\theta} = \frac{1}{2} \bar{x})$$

For Data set B,  $\hat{\theta} = 1,424.4/2 = 712.2$  and  $l(\hat{\theta}) = -179.98$

\* ③ For the gamma dist<sup>n</sup> with unknown parameter:

$$f(x|\alpha, \theta) = \frac{x^{\alpha-1} e^{-x/\theta}}{\Gamma(\alpha) \theta^\alpha}$$

$$\Rightarrow \ln f(x|\alpha, \theta) = (\alpha-1) \ln x - x \theta^{-1} - \ln \Gamma(\alpha) - \alpha \ln \theta$$

$$\therefore l(\alpha, \theta) = \sum_{j=1}^n \ln f_j(x_j|\alpha, \theta)$$

$$\Rightarrow l(\alpha, \theta) = (\alpha-1) \sum_{j=1}^n \ln x_j - \frac{n \bar{x}}{\theta} - n \ln \Gamma(\alpha) - n \alpha \ln \theta \quad (1)$$

To get  $\hat{\alpha}, \hat{\theta}$ , Set  $l'_{\theta}(\alpha, \theta) = 0$  and  $l'_{\alpha}(\alpha, \theta) = 0$  and solving the resulting Eqns

$$l'_{\theta}(\alpha, \theta) = \frac{n \bar{x}}{\theta^2} - \frac{n \alpha}{\theta} = 0 \Rightarrow \frac{\bar{x}}{\theta} = \alpha \quad (2)$$

$$l'_{\alpha}(\alpha, \theta) = \sum_{j=1}^n \ln x_j - \frac{n}{\Gamma'(\alpha)} \text{diff}[\Gamma(\alpha)] - n \ln \theta = 0 \quad (3)$$

The resulting eqns (2) and (3) cannot be solved analytically. Using numerical methods (by using MATLAB or Excel), the estimates are  $\hat{\alpha} = 0.55616$  and  $\hat{\theta} = 2581.1$  and  $l(\hat{\alpha}, \hat{\theta}) = -162.29 \neq$