

5. Suppose that X is a discrete r.v. w/
 $P(X=1) = \theta$ and $P(X=2) = 1-\theta$.
 Three independent obs are made: $X_1=1, X_2=2, X_3=2$.

a) find the method of moments estimate of θ

for discrete r.v., $EX = \sum (x \cdot P(X=x))$
 for this r.v., $EX = 1 \cdot P(X=1) + 2 \cdot P(X=2) = \frac{5}{3}$
 $\frac{5}{3} = \theta + 2 \cdot (1-\theta)$
 $\theta = \frac{1}{3}$

b) what is the likelihood function?

$$L(\theta) = f(x_1, \theta) \cdot f(x_2, \theta) \cdot f(x_3, \theta)$$

$$= \theta \cdot (1-\theta)^2$$

c) what is the maximum likelihood estimate of θ ?

mle $\rightarrow \ln L(\theta) = \ln(\theta(1-\theta)^2)$
 $= \ln \theta + 2 \ln(1-\theta)$
 $\frac{\partial \ln L}{\partial \theta} = \frac{1}{\theta} - \frac{2}{1-\theta}$
 $0 = \frac{1}{\theta} - \frac{2}{1-\theta}$
 $1-\theta = 2\theta$
 $\theta = \frac{1}{3}$

look for critical point

7. Suppose that X follows a geometric distribution:

$$P(X=k) = p(1-p)^{k-1}$$

assume an iid sample size of n

a) find the method of moments estimate of p

EX of geometric = $\sum_{k=1}^{\infty} k \cdot p(1-p)^{k-1} = \frac{1}{p}$
 $\hat{\mu} = \frac{1}{n} \sum_{i=1}^n X_i = \bar{X}$
 $\frac{1}{p} = \bar{X} \Rightarrow \hat{p} = \frac{1}{\bar{X}}$

b) find the mle of p

$$L(p) = \prod_{i=1}^n [p(1-p)^{x_i-1}] = p^n (1-p)^{\sum_{i=1}^n (x_i-1)}$$

$\ln L(p) = n \ln p + [\sum_{i=1}^n (x_i-1)] \ln(1-p)$
 $\frac{\partial \ln L}{\partial p} = \frac{n}{p} - \frac{\sum_{i=1}^n (x_i-1)}{1-p} = 0$
 $\frac{1}{p} = \frac{\sum_{i=1}^n x_i - n}{1-p}$
 $\frac{1}{p} + 1 = \frac{\sum_{i=1}^n x_i}{1-p}$
 $\frac{1}{p} = \frac{\sum_{i=1}^n x_i}{1-p}$
 $\hat{p} = \frac{1}{\bar{X}}$

81. Suppose that X_1, X_2, X_n are iid w/ density function:

$$f(x|\theta) = \begin{cases} e^{-(x-\theta)} & x \geq \theta \\ 0 & \text{otherwise} \end{cases}$$

a) find the method of moments estimate of θ

a) $EX = \int_{\theta}^{\infty} x f(x|\theta) dx = \int_{\theta}^{\infty} x e^{-(x-\theta)} dx$
 $\rightarrow \theta + \int_{\theta}^{\infty} y e^{-y} dy$
 $y = x - \theta \Rightarrow x = y + \theta$
 $dy = dx$
 $= \theta + \int_0^{\infty} (y + \theta) e^{-y} dy = \theta + 1$
 $\hat{\mu}_1 = \hat{\mu}$
 $\theta + 1 = \frac{1}{n} \sum_{i=1}^n X_i = \bar{X}$
 $\hat{\theta} = \bar{X} - 1$

b) find the mle of θ (for what values of θ is the likelihood maximized?)

$$L(\theta) = \prod_{i=1}^n e^{-(x_i-\theta)} = e^{-\sum_{i=1}^n x_i + n\theta} = e^{-\sum x_i} \cdot e^{n\theta}$$

the likelihood function is maximized when θ is maximized as larger θ values will lead to smaller output for $x \geq \theta$.

47. The Pareto dist has been used in economics as a model for a density function of a family's income.

$$f(x|x_0, \theta) = \theta x_0^{\theta} x^{-(\theta+1)}, \quad x > x_0, \theta > 1$$

a) find the method of moments estimate of θ

$EX = \bar{X} = \frac{x_0 \theta}{\theta-1} \Rightarrow x_0 = \frac{\bar{X}(\theta-1)}{\theta}$
 $S^2 = \frac{x_0^2 \theta}{(\theta-1)^2 (\theta-2)} = \frac{x_0^2 \theta}{(\theta-1)(\theta-2)} \Rightarrow x_0 = \frac{S^2 (\theta-1)(\theta-2)}{\bar{X}}$
 $x_0 = \frac{\bar{X}(\theta-1)}{\theta} = \frac{S^2 (\theta-1)(\theta-2)}{\bar{X}}$
 $\frac{\bar{X}^2}{S^2} = \frac{\theta(\theta-1)}{\theta(\theta-2)}$
 $\frac{\bar{X}^2}{S^2} = \frac{\theta-1}{\theta-2}$
 $0 = \theta^2 - 2\theta - (\frac{\bar{X}^2}{S^2})$
 $\theta = \frac{2 \pm \sqrt{4 + 4(\frac{\bar{X}^2}{S^2})}}{2} = 1 \pm \sqrt{1 + (\frac{\bar{X}^2}{S^2})}$
 $\theta > 1$, so $\theta = 1 + \sqrt{1 + (\frac{\bar{X}^2}{S^2})}$
 $x_0 = \frac{\bar{X}(\theta-1)}{\theta}$

b) find the mle of θ

$L(\theta, x_0, x) = \prod_{i=1}^n \theta x_0^{\theta} x_i^{-(\theta+1)} = \theta^n x_0^{n\theta} \prod_{i=1}^n x_i^{-(\theta+1)}$
 $\ln L = n \ln \theta + n \ln x_0 + \sum_{i=1}^n [-(\theta+1) \ln x_i]$
 $\frac{\partial \ln L}{\partial \theta} = \frac{n}{\theta} + n \ln x_0 - \sum_{i=1}^n \ln x_i = 0$
 $\frac{n}{\theta} = \sum_{i=1}^n \ln x_i - n \ln x_0$
 $\frac{1}{\theta} = \frac{\sum_{i=1}^n \ln x_i - n \ln x_0}{n}$

50. Let X_1, \dots, X_n be an iid sample from a Rayleigh distribution w/ param $\theta > 0$

$$f(x|\theta) = \frac{x}{\theta^2} e^{-x^2/(2\theta^2)}, \quad x > 0$$

(All derivations are in 3.4.2)

a) find the method of moments estimate for θ

$EX = \bar{X} = \theta \sqrt{\frac{\pi}{2}}, \theta = \frac{\bar{X}}{\sqrt{\frac{\pi}{2}}}$
 $S^2 = \frac{4-\pi}{2} \theta^2, \theta^2 = S^2 \left(\frac{2}{4-\pi} \right)$
 $\theta = \frac{S \sqrt{2}}{\sqrt{4-\pi}}$
 $\theta = \bar{X} \sqrt{\frac{2}{\pi}} = S \sqrt{\frac{2}{4-\pi}}$

b) find the mle of θ

$L(\theta) = \prod_{i=1}^n \frac{x_i}{\theta^2} e^{-x_i^2/(2\theta^2)} = \frac{1}{\theta^{2n}} e^{-\sum_{i=1}^n x_i^2/(2\theta^2)}$
 $\ln L = \sum_{i=1}^n \ln x_i - 2n \ln \theta - \frac{1}{2\theta^2} \left(\sum_{i=1}^n x_i^2 \right)$
 $\frac{\partial \ln L}{\partial \theta} = 0 = -\frac{2n}{\theta} + \frac{1}{\theta^3} \left(\sum_{i=1}^n x_i^2 \right)$
 $\frac{2n}{\theta} = \frac{1}{\theta^3} \left(\sum_{i=1}^n x_i^2 \right)$
 $\theta^2 = \frac{1}{n} \left(\sum_{i=1}^n x_i^2 \right)$
 $\theta = \sqrt{\frac{1}{n} \left(\sum_{i=1}^n x_i^2 \right)}, \quad \theta > 0$