## Homework 1

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September 25, 2017

1. a. Estimate Maximum Likelihood for

$$f(x) = \frac{x}{\theta^2} exp \frac{-x^2}{2\theta^2}$$

so Maximum Likelihood function (ML) =  $\prod_{1}^{N} \frac{x_i}{\theta^2} exp \frac{-x_i^2}{2\theta^2}$   $\Longrightarrow$  Maximum Log Likelihood function (MLL) =  $\sum_{1}^{N} [\log x_i - 2 \log \theta - \frac{x_i^2}{2\theta^2}]$ 

$$\Longrightarrow$$
 Maximum Log Likelihood function (MLL)

$$\sum_{1}^{N} \left[ \log x_i - 2 \log \theta - \frac{x_i^2}{2\theta^2} \right]$$

To maximize MLL we need to set  $\frac{\partial MLL}{\partial \theta} = 0 \implies \sum_{1}^{N} \left[0 - \frac{2}{\theta} - \frac{x_{i}^{2}}{2} \frac{-2}{\theta^{3}}\right] = 0$   $\implies \sum_{1}^{N} \left[\frac{-2}{\theta} + \frac{x_{i}^{2}}{\theta^{3}}\right] = 0 \implies \frac{1}{\theta} \sum_{1}^{N} \left[2 + \frac{x_{i}^{2}}{\theta^{2}}\right] = 0$   $\implies \sum_{1}^{N} \left[-2 + \frac{x_{i}^{2}}{\theta^{2}}\right] = 0$   $\implies \sum_{1}^{N} -2 + \sum_{1}^{N} \frac{x_{i}^{2}}{\theta^{2}} = 0$   $\implies -2N + \frac{1}{\theta^{2}} \sum_{1}^{N} x_{i}^{2} = 0$   $\implies \frac{1}{\theta^{2}} \sum_{1}^{N} x_{i}^{2} = 2N$   $\implies \frac{\sum_{1}^{N} x_{i}^{2}}{2N} = \theta^{2}$   $\stackrel{\stackrel{?}{=}}{\frac{1}{2}} \sum_{1}^{N} x_{i}^{2} = 0$ 

$$\implies \sum_{1}^{N} \left[ \frac{-2}{\theta} + \frac{x_{i}^{2}}{\theta^{3}} \right] = 0 \implies \frac{1}{\theta} \sum_{1}^{N} \left[ 2 + \frac{x_{i}^{2}}{\theta^{2}} \right] = 0$$

$$\implies \sum_{1}^{N} \left[ -2 + \frac{x_i^2}{\theta^2} \right] = 0$$

$$\implies \sum_{1}^{N} -2 + \sum_{1}^{N} \frac{x_i^2}{\theta^2} = 0$$

$$\implies -2N + \frac{1}{\theta^2} \sum_{i=1}^{N} x_i^2 = 0$$

$$\implies \frac{1}{42} \sum_{i=1}^{N} x_i^2 = 2N$$

$$\implies \frac{\sum_{1}^{N} x_{i}^{2}}{2N} = \theta^{2}$$

$$\implies \hat{\theta} = \sqrt{\frac{\sum_{1}^{N} x_{i}^{2}}{2N}}$$

c. 
$$f(x) = \frac{1}{\theta} \forall 0 <= x <= \theta$$

so Maximum Likelihood (ML) =  $\prod_1^N \frac{1}{\theta}$   $\Longrightarrow$   $ML = \frac{1}{\theta^N}$ 

$$\implies ML = \frac{1}{\theta^N}$$

To maximize the above likelihood we need to minimize  $\theta$ 

But we also have a constraint that  $0 \le x \le \theta$ 

so 
$$\hat{\theta} = max[x_i]$$