

1. Answer problems 3 and 13 from chapter 2 of your textbook. On problem 13, also obtain the expected value and the variance of the ML estimator. Recall that the density associated with a random variable with exponential distribution is given by

$$f(x) = \frac{1}{\theta} \exp(-x/\theta) \text{ for } \theta, x > 0.$$

**Answers:** 3. By definition,

$$\hat{\sigma}_{ML}^2 = \operatorname{argmax}_{\sigma^2 \in (0, \infty)} \frac{1}{n} \sum_{i=1}^n \log \left( \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left( -\frac{1}{2} \frac{(X_i - \mu)^2}{\sigma^2} \right) \right) \equiv \operatorname{argmax}_{\sigma^2 \in (0, \infty)} \ell_n(\sigma^2).$$

Now,

$$\frac{d}{d\sigma^2} \ell_n(\sigma^2) = \frac{1}{n} \sum_{i=1}^n \left( -\frac{1}{2\sigma^2} + \frac{1}{2(\sigma^2)^2} \frac{(X_i - \mu)^2}{\sigma^2} \right)$$

and solving for  $\hat{\sigma}_{ML}^2$  such that  $\frac{d}{d\sigma^2} \ell_n(\hat{\sigma}_{ML}^2) = 0$  gives  $\hat{\sigma}_{ML}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2$ .

**Answers:** 13. The density associated with a random variable with exponential distribution is given by

$$f(x) = \frac{1}{\theta} \exp(-x/\theta) \text{ for } \theta, x > 0.$$

Then, the log-likelihood function is given by

$$\ell_n(\theta) = \operatorname{argmax}_{\theta \in (0, \infty)} \frac{1}{n} \sum_{i=1}^n \log \left( \frac{1}{\theta} \exp(-X_i/\theta) \right).$$

Taking a derivative with respect to  $\theta$  and solving for  $\hat{\theta}_{ML}$  such that  $\frac{d}{d\theta} \ell_n(\hat{\theta}_{ML}) = 0$ , we have

$$\frac{d}{d\theta} \ell_n(\hat{\theta}_{ML}) = \frac{1}{n} \sum_{i=1}^n \left( -\frac{1}{\hat{\theta}_{ML}} + \frac{1}{\hat{\theta}_{ML}^2} X_i \right) = 0$$

which implies  $\hat{\theta}_{ML} = \frac{1}{n} \sum_{i=1}^n X_i$ .

The expected value of  $\hat{\theta}_{ML}$  is  $E(\hat{\theta}_{ML}) = E\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \theta$ . The variance is

$$V(\hat{\theta}_{ML}) = V\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} n V(X_i) = \theta^2/n.$$

2. In class we defined the empirical distribution function  $F_n(r)$  associated with a random sample  $\{r_t\}_{t=1}^n$ ; a) Obtain an asymptotically valid  $\alpha = 0.95$  confidence interval for  $F_r(r)$  the distribution of  $r_t$ ; b) Use the APPL.mat data in the class website to test the hypothesis that log-return on Apple stock is normally distributed; c) Test the hypothesis that log-return on Apple stock is comes from a location-scale Student-t distribution.

**Answer:** a) As derived in class we have that

$$E(F_n(r)) = F_r(r) \text{ and } V(F_n(r)) = \frac{1}{n} (1 - F_r(r)) F_r(r)$$

Hence, by Lévy's CLT

$$\frac{\sqrt{n}}{(1 - F_r(r))F_r(r)} (F_n(r) - F_r(r)) \rightarrow N(0, 1).$$

Hence,

$$P\left(-1.96 \leq \frac{\sqrt{n}}{(1 - F_r(r))F_r(r)} (F_n(r) - F_r(r)) \leq 1.96\right) = 0.95$$

b) Use MLESP500\_KS.m with Apple data.

**Answer:** Directly from the code provided in the website.

c) Use MLE\_SP500\_student\_t.m with Apple data.

**Answer:** Directly from the code provided in the website.

3. Use the data set AAPL.mat to obtain daily log-returns on Apple stock. Assuming that these data are:

- (a) independently drawn and identically distributed as  $Student-t(\mu, \sigma^2, \nu)$ . Estimate the parameters of this distribution by Maximum Likelihood using MATLAB.

**Answers:** Use codes MLE\_SP500.m and MLE\_SP500\_student\_t.m for items (a) and (b), respectively changing the data set used to AAPL.mat.

4. In class we considered, and used, two estimation procedures - the method of moments and the method of maximum likelihood. Suppose that you are interested in a random variable  $X$  that has the following density

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } x \in [a, b] \\ 0 & \text{if } x \notin [a, b] \end{cases}$$

where  $a, b \in \mathbb{R}$  with  $b > a$ . The values of  $a$  and  $b$  are unknown and must be estimated using a random sample  $\{X_i\}_{i=1}^n$ . Obtain the method of moments estimators for  $a$  and  $b$ . Suppose that instead of the method of moments estimator for  $b$  you propose the following estimator,

$$\tilde{b}_n = \max_{1 \leq i \leq n} \{X_i\}.$$

What is the probability that  $|\tilde{b}_n - b|$  will exceed a certain  $\epsilon > 0$ ?

**Answer:** If the random variable  $X$  has density given by  $f$ , then  $E(X) = \frac{a+b}{2}$  and  $V(X) = \frac{(b-a)^2}{12}$ . we use these two moments to obtain estimators for  $a$  and  $b$ . Let  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  and  $s^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$  and define the following equations

$$\bar{X} = \frac{\hat{a}_n + \hat{b}_n}{2} \tag{1}$$

and

$$s^2 = \frac{(\hat{b}_n - \hat{a}_n)^2}{12}. \tag{2}$$

Solving equations (1) and (2) for  $\hat{a}_n$  and  $\hat{b}_n$  we obtain,  $\hat{a}_n = \bar{X} - s\sqrt{3}$  and  $\hat{b}_n = \bar{X} + s\sqrt{3}$ .

$P(|\tilde{b}_n - b| < \epsilon) = P(-(\tilde{b}_n - b) < \epsilon) = P(\tilde{b}_n > b - \epsilon) = 1 - P(\tilde{b}_n \leq b - \epsilon)$ . But

$$\begin{aligned} P(\max_{1 \leq i \leq n} \{X_i\} \leq x) &= P(X_1 \leq x, \dots, X_n \leq x) \\ &= P(X_1 \leq x)P(X_2 \leq x) \cdots P(X_n \leq x), \text{ by independence} \\ &= F(x)^n \text{ by the fact that the distribution is identical for all } X_i \end{aligned}$$

Hence,

$$1 - P(\tilde{b}_n \leq b - \epsilon) = P(|\tilde{b}_n - b| < \epsilon) = 1 - P(|\tilde{b}_n - b| > \epsilon)$$

which implies  $P(|\tilde{b}_n - b| > \epsilon) = F^n(b - \epsilon)$ . Since  $F$  takes values between 0 and 1, as  $n \rightarrow \infty$  we have  $P(|\tilde{b}_n - b| > \epsilon) \rightarrow 0$ .