## PROBABILITY AND STATISTICS Exercise sheet 11

**MC 11.1.** Let  $n \in \mathbb{N}$  and let  $X_1, \ldots, X_n$  be i.i.d. Bernoulli(p) random variables for some  $p \in (0, 1)$ . (Exactly one answer is correct in each question.)

- (i) What is the maximum likelihood estimator of p?
  - (a)  $T_{ML} = \overline{X}_n$ .
  - (b)  $T_{ML} = 1/\overline{X}_n$ .
  - (c)  $T_{ML} = \max\{X_i : i \in \{1, \dots, n\}\}.$
  - (d)  $T_{ML} = X_n$ .
- (ii) Is the maximum likelihood estimator unbiased?
  - (a) Yes.
  - (b) No.
- (iii) What is the distribution of the random variable  $n \times T_{ML}$ ?
  - (a)  $n \times T_{ML} \sim \text{Binom}(n, p)$ .
  - (b)  $n \times T_{ML} \sim \text{Geom}(p)$ .
  - (c)  $n \times T_{ML} \sim \text{Binom}(n, p/n).$
  - (d)  $n \times T_{ML} \sim \text{Bernoulli}(np).$

**MC 11.2.** Let  $X_1, \ldots, X_n$ , for  $n \in \mathbb{N}$  be i.i.d. random variables with  $X_i \sim \mathcal{U}([-\theta, \theta])$  for some unknown  $\theta \in (0, \infty)$ . What is the maximum likelihood estimator for  $\theta$ ? (Exactly one answer is correct.)

- (a)  $T_{ML} = \frac{2}{n} \sum_{i=1}^{n} X_i \mathbf{1}_{\{X_i > 0\}}.$
- (b)  $T_{ML} = \frac{1}{n} \sum_{i=1}^{n} 2X_i.$
- (c)  $T_{ML} = \max\{|X_i| : i \in \{1, \dots, n\}\}.$
- (d) The maximum likelihood estimator does not exist.

**MC 11.3.** Let X be a random variable with well-defined positive variance. Find  $a \in \mathbb{R}$  and b > 0 such that for Z := (X - a)/b, it holds that  $\mathbb{E}[Z] = 0$  and  $\operatorname{Var}[Z] = 1$ . (Exactly one answer is correct.)

- (a)  $a = \mathbb{E}[X], b = \operatorname{Var}[X].$
- (b)  $a = \mathbb{E}[X], b = \sqrt{\operatorname{Var}[X]}.$
- (c)  $a = \mathbb{E}[X^2], b = \operatorname{Var}[X].$
- (d)  $a = (\mathbb{E}[X])^2, b = \operatorname{Var}[X].$

**Exercise 11.4.** Let  $X_1, \ldots, X_n$  be independent random variables with  $X_i \sim \mathcal{N}(\theta \alpha_i, 1)$ , where  $\alpha_i \neq 0$  are known parameters, but  $\theta \in \mathbb{R}$  is unknown. Find the maximum likelihood estimator for  $\theta$ .

**Exercise 11.5.** In a sawmill, the sawn timber of a certain sorting class undergoes quality control. Each production day, a sample of ten boards is taken, and each board is tested for its stiffness. Based on experience, it can be assumed that the stiffness of a board is normally distributed with a known standard deviation  $\sigma = 1430 \text{ MPa}^{1}$ .

- (a) Derive the formula for the 95% confidence interval for  $\mu$  after 15 production days. You can assume that the samples are mutually independent.
- (b) Compute the realized confidence interval from (a) for a sample mean of  $\overline{x} = 11,000$  MPa (after 15 production days).
- (c) How many samples would be required to ensure that the width of the confidence interval is less than 200 MPa?

**Exercise 11.6.** Let  $\Theta = [1/2, 1]$ . We consider the model family  $(\mathbb{P}_{\theta})_{\theta \in \Theta}$ , where  $X_1, \ldots, X_n$  are independent and identically distributed under  $\mathbb{P}_{\theta}$  with  $X_1 \sim \text{Geom}(\theta)$ . The maximum likelihood estimator for  $\theta$  is given by (check!)

$$T_{ML} = \frac{n}{\sum_{i=1}^{n} X_i}.$$

Find an approximate 95% confidence interval for  $\theta$ .

**Hint:** For all  $\theta \in [1/2, 1]$ , we have  $\frac{\sqrt{1-\theta}}{\theta} \leq \sqrt{2}$ . You can use this to simplify the expression. Note that enlarging the interval only increases the probability of coverage. However, for  $\theta$  close to 1, the interval might be unnecessarily wide.

Hint 2: Recall that we know that

$$\mathbb{E}_{\theta}[X_1] = \frac{1}{\theta}$$
 and  $\operatorname{Var}_{\theta}[X_1] = \frac{1-\theta}{\theta^2}$ .

**Remark:** In practice, if the variance is unknown, we typically replace it with an estimate. Either we use the generic estimator

$$\hat{\sigma}_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \overline{X}_n)^2,$$

or, since in this case  $\operatorname{Var}[X_1] = \frac{1-\theta}{\theta^2}$  and we have an estimator for  $\theta$ , we can also use

$$\tilde{\sigma}_n^2 = \frac{1 - T_{ML}}{T_{ML}^2}.$$

As both estimators are consistent in this case, we still have by Slutsky's theorem that

$$\frac{\sum_{i=1}^{n} X_i - \frac{n}{\theta}}{\sqrt{n \hat{\sigma}_n^2}} \approx \mathcal{N}(0, 1) \quad \text{and} \quad \frac{\sum_{i=1}^{n} X_i - \frac{n}{\theta}}{\sqrt{n \tilde{\sigma}_n^2}} \approx \mathcal{N}(0, 1).$$

**Exercise 11.7.** To estimate the number N of trout in a lake, the following procedure is used (capture-recapture method): In a first step, 500 trout are caught, marked, and released back into the lake. In a second step, 200 trout are caught again, and the number X of marked trout is recorded.

<sup>&</sup>lt;sup>1</sup>The pascal is a derived SI unit of pressure and mechanical stress. It is named after Blaise Pascal and defined as follows: 1 Pa =  $1 \text{ kg} \times \text{m}^{-1} \times \text{s}^{-2} = 1 \text{ N} \times \text{m}^{-2}$ . One pascal is the pressure exerted by a force of one newton on an area of one square meter. 1 MPa =  $10^6$  Pa = 10 bar.

- (a) X is modeled as a binomial random variable,  $X \sim \text{Binom}(n, \theta)$ , where  $\theta$  denotes the probability that a fish caught in the second step is marked. What is the value of n? What is the value of the parameter  $\theta$  if the total number of trout in the lake is N = 2000 or N = 5000?
- (b) The actual observed value for X is 40. Give a reasonable estimate for the parameter  $\theta$ , and derive from it an estimate for the total number N of trout in the lake.
- (c) Determine an approximate 95% confidence interval for  $\theta$ , and from that, an approximate 95% confidence interval for N.

0.5	0.75	0.9	0.95	0.975	0.99	0.995	0.999
0	0.6745	1.2816	1.6449	1.9600	2.3263	2.5758	3.0902

## Quantile table for the standard normal distribution

For instance,  $\Phi^{-1}(0.9) = 1.2816$ , where  $\Phi$  is the distribution function of  $\mathcal{N}(0,1)$ .

0.000.010.02 0.030.040.050.060.070.08 0.090.5040 0.50800.5120 0.5199 0.5239 0.5279 0.5319 0.5359 0.00.50000.51600.10.53980.54380.54780.55170.55570.5596 0.56360.56750.57140.57530.20.57930.58320.58710.59100.59480.5987 0.6026 0.6064 0.61030.61410.30.62170.62550.6293 0.63310.6368 0.64060.64430.64800.65170.61790.6628 0.67000.67720.40.65540.65910.6664 0.67360.6808 0.68440.68790.69500.69850.7019 0.70540.71230.71900.50.69150.7088 0.71570.72240.60.72570.72910.73240.73570.73890.74220.74540.74860.75170.75490.70.75800.76110.76420.76730.77040.77340.77640.77940.78230.78520.79950.80.78810.79100.79390.7967 0.80230.80510.80780.81060.8133 0.90.81590.81860.8212 0.82380.82640.82890.83150.83400.83650.8389 1.00.84130.84380.84610.84850.85080.85310.85540.85770.85990.86211.1 0.86430.86650.8686 0.8708 0.8729 0.8749 0.87700.8790 0.8810 0.8830 1.20.88490.88690.88880.89070.89250.89440.8962 0.89800.8997 0.90151.30.90320.9049 0.9066 0.90820.9099 0.91150.91310.91470.91620.91771.40.9192 0.9207 0.92220.9236 0.92510.92650.9279 0.9292 0.93060.9319 0.93821.50.93320.93450.93570.93700.93940.94060.94180.94290.94411.60.94520.94630.94740.94840.94950.95050.95150.95250.95350.95451.70.95540.95640.95730.95820.95910.9599 0.9608 0.9616 0.96250.9633 1.80.96410.96490.96560.96640.96710.9678 0.96860.96930.96990.9706 1.90.97130.97190.97260.97320.97380.9744 0.97500.9756 0.9761 0.9767 2.00.97720.97780.97830.97880.97930.9798 0.98030.98080.98120.98172.10.98210.98260.98340.98380.98460.98540.98300.98420.98500.98572.20.9868 0.98750.98810.98900.98610.98640.9871 0.98780.98840.98872.30.98930.98960.98980.9901 0.99040.9906 0.99090.99110.99130.99162.40.99180.99200.9922 0.9925 0.99270.9929 0.99310.99320.99340.9936 2.50.9938 0.99400.99410.99430.99450.9946 0.99480.99490.99510.99522.60.99530.99550.9956 0.9957 0.9959 0.9960 0.9961 0.99620.9963 0.99640.9969 2.70.99650.9966 0.99670.9968 0.9970 0.99710.99720.99730.99742.80.99740.99750.9976 0.9977 0.99770.9978 0.99790.99790.99800.99812.90.99810.99820.99820.99830.99840.99840.99850.9985 0.9986 0.99863.00.99870.9987 0.9987 0.9988 0.9988 0.9989 0.9989 0.99890.99900.9990

## Table of standard normal distribution

For instance,  $\mathbb{P}[Z \le 1.96] = 0.975$ .