

## Statistics: Tutorial sheet 2

### Mandatory exercises

**Exercise 1.** For each of the following distributions, indicate whether it is a member of the exponential family and, if yes, provide expressions for each component inside the definition of the exponential family.

- a. Exponential( $\lambda$ ):  $f(x|\lambda) = \lambda e^{-\lambda x}$ ;  $0 \leq x < \infty$ ,  $\lambda > 0$ ,
- b. Gamma( $\alpha, \beta$ ):  $f(x|\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}$ ;  $0 \leq x < \infty$ ,  $\alpha, \beta > 0$ ,
- c. Uniform( $0, \theta$ ):  $f(x|\theta) = \frac{1}{\theta}$ ;  $0 \leq x \leq \theta$ ,  $\theta > 0$ ,
- d. Poisson( $\lambda$ ):  $f(x|\lambda) = \frac{e^{-\lambda} \lambda^x}{x!}$ ;  $x = 0, 1, \dots$ ,  $\lambda > 0$ .

**Exercise 2.** Let  $g(x|\mu, \sigma)$  with cdf  $G(x|\mu, \sigma)$  be a member of the location-scale family defined by  $g(x)$  with cdf  $G(x)$ . Prove the following general results:

- a.  $G(x|\mu, \sigma) = G\left(\frac{x-\mu}{\sigma}\right)$ ,
- b.  $G^{-1}(x|\mu, \sigma) = \mu + \sigma G^{-1}(x)$ .

### Practice exercises

**Exercise 1.** A student has gotten very excited from the histogram examples and has decided to test them out himself. He simulates a data set of fifty observations from the standard normal distribution. Then he creates a histogram and plots the standard normal pdf in one figure. The result is shown in Figure 1.

- a. Recall that for  $y \in (a_{j-1}, a_j]$  we defined the histogram function as

$$h_n(y) = \sum_{j=1}^m \mathbb{1}_{\{a_{j-1} < y \leq a_j\}} \left( \sum_{i=1}^n \mathbb{1}_{\{a_{j-1} < x_i \leq a_j\}} \right).$$

Formally solve the integral

$$\int_{\mathbb{R}} h_n(y) dy.$$

- b. Explain what mistake the student has made and provide a solution.

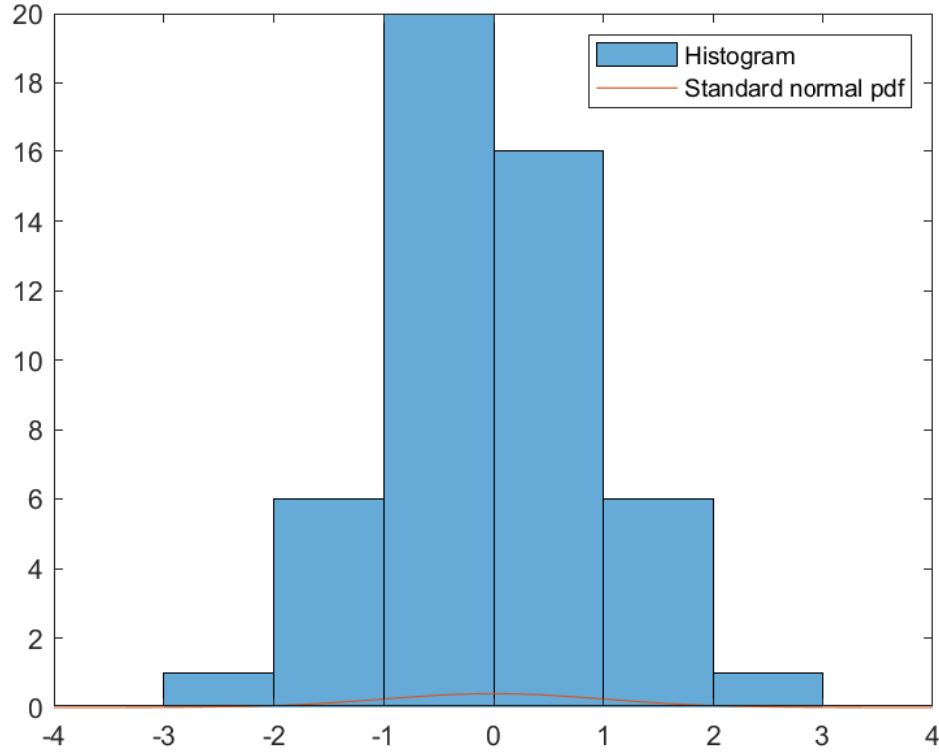


Figure 1: Estimated histogram

**Exercise 2.** Let  $X_1, \dots, X_n$  be a random sample with population  $g(x)$  and let  $Y \sim g(x)$  be independent of each  $X_i$ .

- Prove that  $P(Y \leq X_{(1)}) = \frac{1}{n+1}$ .
- Prove for  $k \in \{2, 3, \dots, n\}$  that we have  $P(X_{(k-1)} < Y \leq X_{(k)}) = \frac{1}{n+1}$ .
- Use this to show that  $P(Y \leq X_{(k)}) = \frac{k}{n+1}$ .

**Exercise 3.** In this exercise we study a new distribution called the *Logistic distribution* which for  $\theta = (\mu, \sigma)$  with  $\mu \in \mathbb{R}$  and  $\sigma > 0$  has cdf

$$G(x|\theta) = \frac{1}{1 + e^{-(x-\mu)/\sigma}} \quad \text{if } x \in \mathbb{R}.$$

The logistic distribution has become very popular in machine learning due to the sharp S shape and easy differentiability of its cdf. This makes it one of the most common distributions to model probabilities of outcomes as in logistic regression<sup>1</sup> or feed forward neural networks.<sup>2</sup>

- Derive the location-scale family of the Logistic(0, 1) distribution.

<sup>1</sup>[https://en.wikipedia.org/wiki/Logistic\\_regression](https://en.wikipedia.org/wiki/Logistic_regression)

<sup>2</sup>[https://en.wikipedia.org/wiki/Feedforward\\_neural\\_network](https://en.wikipedia.org/wiki/Feedforward_neural_network)

- b. Verify that  $G(x|\mu, \sigma)^{-1} = \mu + \sigma G(x|0, 1)^{-1}$ .

**Exercise 4.** Suppose that a marine biologist wants to know the number of fish  $N_0$  that live in a lake. The water is not very clear, so she cannot count them from a helicopter or use other such practical methods. Instead she comes up with a different approach. She starts fishing in the lake, catches  $r$  different fish, puts a mark on each of them and then throws them back into the water. One week later she comes back and performs an experiment by catching a fish, writing down if its marked or not and throwing it back. After repeating this for a total of  $n$  times she has obtained a dataset  $\mathbf{x} = (x_1, \dots, x_n)$ .

- a. Assume that the random variables  $X_1, \dots, X_n$  are independent. Formulate a statistical model for  $\mathbf{X} = (X_1, \dots, X_n)$ . What would be an intuitive estimator for  $N_0$ ?
- b. Is the independence assumption reasonable here?