## University of Wrocław: Data Science

Theoretical Foundations of Large Data Sets, List 3

## **Definitions**

$$V_i(t) = \mathbf{1}_{\{p_i \le t\}}, \qquad F_n(t) = \frac{\sum V_i}{n}.$$

• Higher Criticism Test (Tukey 1976):

$$HC^* = \max_{1/n < t < 1/2} \sqrt{n} \frac{F_n(t) - t}{\sqrt{t(1-t)}}$$

• Modification by Stepanova and Pavlenko (2014)

$$HC_{mod} = \max_{0 < t < 1} \sqrt{n} \frac{F_n(t) - t}{\sqrt{t(1 - t)q(t)}}, \qquad q(t) = \log \log \frac{1}{t(1 - t)}.$$

- 1. For  $n \in \{5000; 50000\}$  estimate the probability of the type I error for  $HC_{mod}$  using the asymptotic critical value for 0.05 significance test  $C_{crit} = 4.14$ .
- 2. For n = 5000 estimate critical values of both Higher-Criticism tests at the significance level  $\alpha = 0.05$ .
- 3. Let n = 5000 and

a) 
$$\mu_1 = 1.2\sqrt{2\log n}, \mu_2 = \ldots = \mu_n = 0;$$

b) 
$$\mu_1 = \ldots = \mu_{100} = 1.02\sqrt{2\log\left(\frac{n}{200}\right)}, \mu_{101} = \ldots = \mu_n = 0;$$

c) 
$$\mu_1 = \dots = \mu_{1000} = 1.002\sqrt{2\log\left(\frac{n}{2000}\right)}, \mu_{1001} = \dots = \mu_n = 0;$$

Use the above settings to compare the power of the following tests and summarize the results:

- Higher-Criticism,
- modified Higher-Criticism,
- Bonferroni,
- chi-square,

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- Fisher,
- Kolmogorov-Smirnov (ks.test),
- Anderson-Darling (ad.test {goftest}).
- 4. Consider the sparse mixture model

$$f(\mu) = (1 - \varepsilon)\delta_0 + \varepsilon \delta_\mu$$

with  $\varepsilon = n^{-\beta}$  and  $\mu = \sqrt{2r \log n}$ .

For each of the settings  $\beta = \{0.6; 0.8\}, r = \{0.1; 0.4\}$  and  $n = \{5000; 50000\}$ :

- a) simulate the critical values for the Neyman-Pearson test in the sparse mixture;
- b) Compare the power of the Neyman-Pearson test to the power of both versions of HC, Bonferroni, Fisher and chi-square.

Summarize the results referring to the theory learned in class.

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