### Normal Distribution

Andrey Chinnov, Sebastian Honermann, Carlos Zydorek

Case Studies "Data Analytics"

### Outline

- Introduction
  - Normality as a requirement for statistical methods
  - Data Set Overview
- Normality Testing
  - Graphical Methods for Normality Testing
    - ★ Q-Q-Plots
    - ★ Chi-Square Plot
  - Quantitative Methods for Normality Testing
    - ★ Shapiro-Wilk Test
    - ★ Pearson's Chi-Squared Test
    - ★ Kolmogorov-Smirnov Test
- Transformation to Normality
  - ▶ Box-Cox Transformation
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## Graphical Methods for Normality Testing Q-Q-Plots

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## Quantitative Methods for Normality Testing Shapiro-Wilk Test

### Quantitative Methods for Normality Testing Pearson's Chi-Squared Test

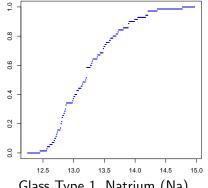
Let  $x = (x_1, x_2, \dots, x_n)$  be a sample of unknown distribution  $\mathbb{P}$ .

### Definition

$$F_n(x) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{x_i \le x\}}(x)$$

- empirical c. d. f., where

$$\mathbb{1}_{\{x_i \le x\}}(x) = \begin{cases} 1 & \text{if } x_i \le x \\ 0 & \text{otherwise.} \end{cases}$$



Glass Type 1, Natrium (Na)

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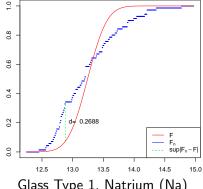
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F(x) - theoretical normal c. d. f. with

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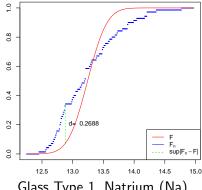
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$$d = \sup_{x \in \mathbb{R}} |F_n(x) - F(x)|$$

- distance between them.



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 $\bullet \ \forall t > 0$ :

$$P(D_n \le t) \xrightarrow[n \to \infty]{} H(t) = 1 - 2\sum_{i=1}^{\infty} (-1)^{i-1} \exp^{-2i^2t^2}$$

The KS test uses the decision rule

$$\delta = \left\{ \begin{array}{ll} H_0 & : & D_n \le c \\ H_1 & : & D_n > c \end{array} \right.,$$

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$$\implies c \approx H_{1-\alpha}$$

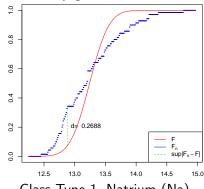
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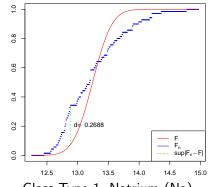
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#### Example:

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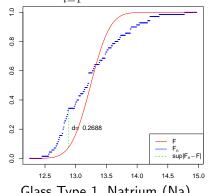
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- n = 70
- $D_n = \sqrt{n} \sup |F_n F| = 2.2493$
- $\alpha = 0.01$

$$\implies c = H_{1-\alpha} = 1.6276$$



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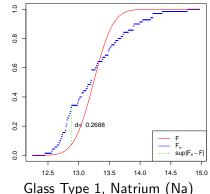
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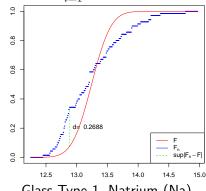
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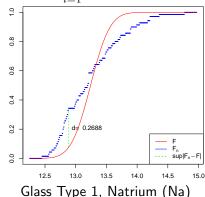
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- $\bullet \implies \mathbb{P} \neq \mathbb{P}_0$
- → data not normally distributed!!!



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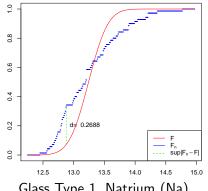
$$KS(\mu, \sigma) = \sup_{x \in \mathbb{R}} |F_n(x) - F(x, \mu, \sigma)| \to \min.$$

KS test is improved by solving the following optimization problem

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Initial vector of parameters

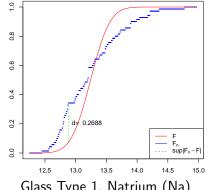
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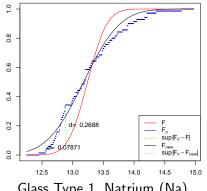
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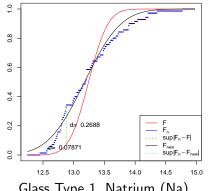
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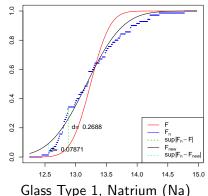
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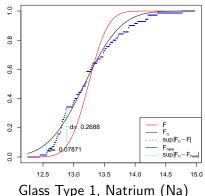
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- $D_n < c \implies H_0$  accepted



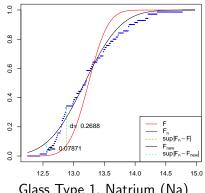
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