### Normal Distribution

Andrey Chinnov, Sebastian Honermann, Carlos Zydorek

### 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
  - Transformation Results Testing
- Summary



### Normality as a requirement for statistical methods

Introduction

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### Graphical Methods for Normality Testing Q-Q-Plots

Normality Testing •0000000000

#### Theretical:

$$p_{(j)} = \frac{j - \frac{1}{2}}{n}$$

$$q_{(j)} = \Phi^{-1}(p_{(j)})$$

## Graphical Methods for Normality Testing Q-Q-Plots

3

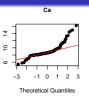
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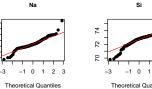
$$p_{(j)} = \frac{j - \frac{1}{2}}{n}$$

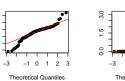
$$q_{(j)} = \Phi^{-1}(p_{(j)})$$













Ва







0.2

0.0



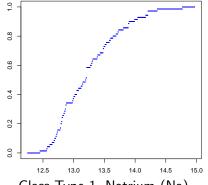
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### Definition

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

- 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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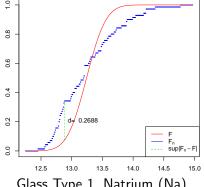
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$$\bar{x} = \frac{1}{n} \sum_{i} x_i, \quad \sigma_x^2 = \frac{1}{n} (x_i - \bar{x})^2$$



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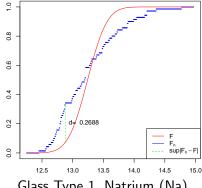
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$$d = \sup_{x \in \mathbb{R}} |F_n(x) - F(x)|$$

- 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

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

where c - critical value

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

The KS test uses the decision rule for a given significance level  $\alpha$ 

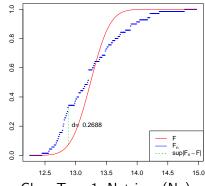
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$$n = 70$$

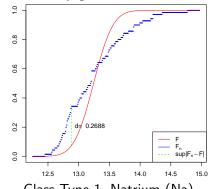
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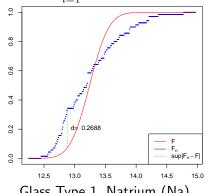
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$$\implies c = H_{1-\alpha} = 1.6276$$



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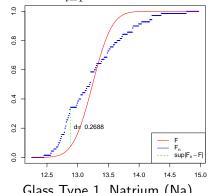
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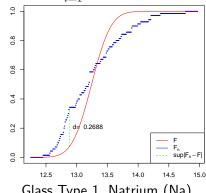
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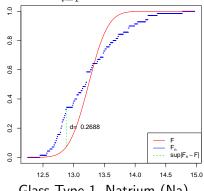
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- → data not normally distributed!!!



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KS test is improved by solving the following optimization problem

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

R code used:

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```
R code used:
```

c(mean(dat), var(dat))

[1] 13.2422857 0.2493019

#optim is a predifined R function in stats package #defalut method of optimization is Nelder and Mead result = optim(c(mean(dat), var(dat)), KS) result\$par

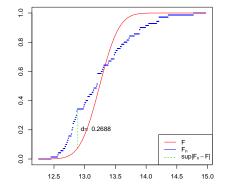
[1] 13.1769501 0.4682486

result\$value

[1] 0.07870673

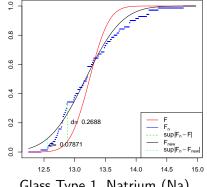
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- Initial vector of parameters  $\mu = 13.2423$ ,  $\sigma^2 = 0.2493$
- Optimized vector of parameters  $\hat{\mu} = 13.1770, \quad \hat{\sigma}^2 = 0.4682$



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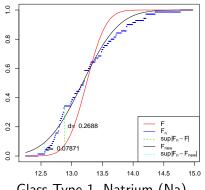
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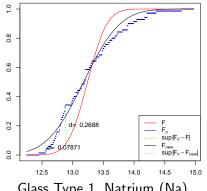
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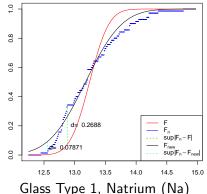
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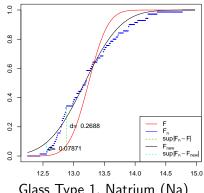
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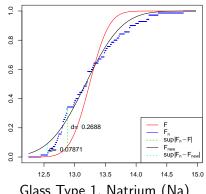


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- ⇒ data normally distributed!



Glass Type 1, Natrium (Na)

variable	test statistic	sig. level	critical value	p-value	rejected
RI	1.34	0.01	1.63	0.0561963016778131	no
Na	0.87	0.01	1.63	0.43825271603342	no
Mg	2.94	0.01	1.63	6.18457917100912e-08	yes
Al	0.84	0.01	1.63	0.474757887353829	no
Si	0.96	0.01	1.63	0.314710019077325	no
K	2.14	0.01	1.63	0.000212776619708754	yes
Ca	1.33	0.01	1.63	0.057710602872685	no
Ba	2.60	0.01	1.63	2.75476085742632e-06	yes
Fe	4.68	0.01	1.63	< 1.0e-15	yes

Results of Improved KS test on the whole data set:

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Al	0.84	0.01	1.63	0.474757887353829	no
Si	0.96	0.01	1.63	0.314710019077325	no
K	2.14	0.01	1.63	0.000212776619708754	yes
Ca	1.33	0.01	1.63	0.057710602872685	no
Ba	2.60	0.01	1.63	2.75476085742632e-06	yes
Fe	4.68	0.01	1.63	< 1.0e-15	yes

- 5 variables are normaly distributed (RI,Na,AI,Si,Ca)
- 4 variables are not (Mg,K,Ba,Fe)
- The best statistics test value for Al
- The worst statistic test value for Fe

## Quantitative Methods for Normality Testing

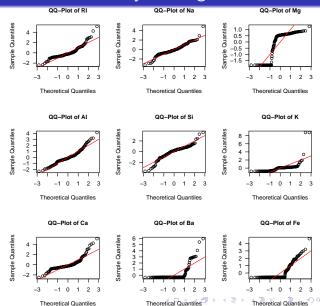
#### **Test Results:**

variable	rejected
RI	no
Na	no
Mg	yes
ΑI	no
Si	no
K	yes
Ca	no
Ba	yes
Fe	yes

## Quantitative Methods for Normality Testing

#### Test Results:

variable	rejected
RI	no
Na	no
Mg	yes
ΑI	no
Si	no
K	yes
Ca	no
Ba	yes
Fe	yes



Theoretical Quantiles

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  - 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
  - Transformation Results Testing
- Summary



Transformation to Normality

### Box-Cox Transformation

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Summary

## Summary

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