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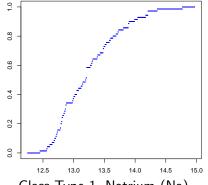
Let $x = (x_1, x_2, \dots, x_n)$ be a sample of unknown distribution \mathbb{P} .

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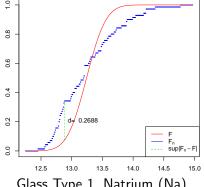
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$$\bar{x} = \frac{1}{n} \sum_{i} x_i, \quad \sigma_x^2 = \frac{1}{n} (x_i - \bar{x})^2$$



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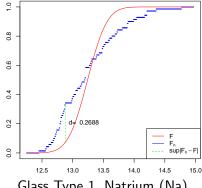
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$$d = \sup_{x \in \mathbb{R}} |F_n(x) - F(x)|$$

- distance between them.



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 \bullet $\forall t>0$:

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

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The KS test uses the decision rule

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

The KS test uses the decision rule for a given significance level α

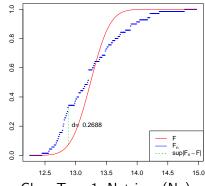
$$\delta = \begin{cases} H_0 : D_n \le H_{1-\alpha} \\ H_1 : D_n > H_{1-\alpha} \end{cases}, \quad H(t) = 1 - 2\sum_{i=1}^{\infty} (-1)^{i-1} \exp^{-2i^2 t^2}$$

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$$n = 70$$

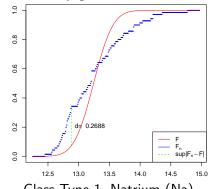
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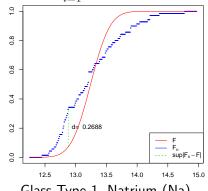
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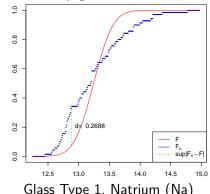
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• $D_n > c \implies H_0$ rejected



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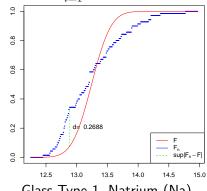
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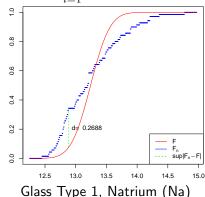
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- $\bullet \implies \mathbb{P} \neq \mathbb{P}_0$
- → data not normally distributed!!!



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

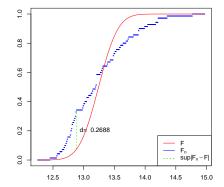
result\$value

[1] 0.07870673

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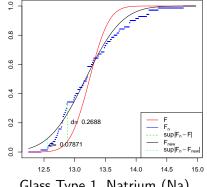
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- Initial vector of parameters $\mu = 13.2423$, $\sigma^2 = 0.2493$
- Optimized vector of parameters $\hat{u} = 13.1770$. $\hat{\sigma}^2 = 0.4682$



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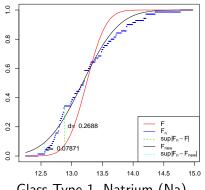
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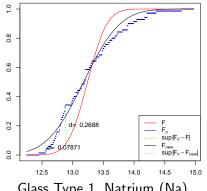
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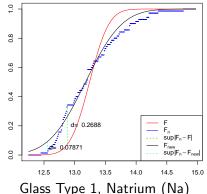
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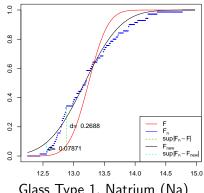
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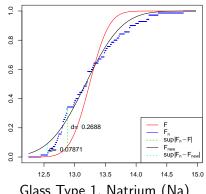


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- $\bullet \implies \mathbb{P} = \mathbb{P}_0$
- ⇒ data normally distributed!



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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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Quantitative Methods for Normality Testing

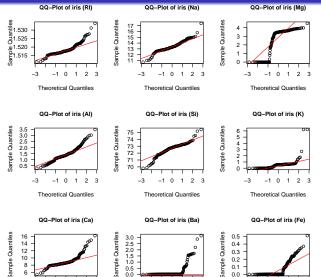
Test Results:

variable	rejected
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Quantitative Methods for Normality Testing

Test Results:

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2 Theoretical Quantiles

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 - ★ Kolmogorov-Smirnov Test
- Transformation to Normality
 - Box-Cox Transformation
 - Transformation Results Testing
- Summary



Transformation to Normality

Box-Cox Transformation

Outline

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 - Normality as a requirement for statistical methods
 - Data Set Overview
- Normality Testing
 - Graphical Methods for Normality Testing
 - ★ Q-Q-Plots
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 - Quantitative Methods for Normality Testing
 - ★ Shapiro-Wilk Test
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Summary