

Tarigan Statistical Consulting & Coaching

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Doctoral Program in Computer Science of the Universities of Fribourg, Geneva, Lausanne, Neuchâtel, Bern and the EPFL



Hands-on Data Analysis with R University of Neuchatel, 10 May 2016

Welcome & Road Map

Bernadetta Tarigan, Dr. sc. ETHZ





| 08:45 – 12:45 | Welcome and Road map Linear & Piecewise Regression (Hands on <i>critic</i> data) GLM & Feature Selection (Hands on <i>bug data</i>) |
|---------------|--|
| 12:45 - 14:15 | Lunch |
| 14:15 – 18:00 | Continue with <i>bug data</i> Two-Way ANOVA (Hands on variabletype data) Wrap up and Evaluation |



When coffee breaks? In between!



Who is talking to you?

| | TEACHING/TUTORING | RESEARCH | COACHING | Consulting & Data Analysis |
|---|--|--|---|--|
| 2010-2016 ETHZ-CH Self- employed (ZH) | | <u>www.mis.ethz.ch</u> <u>www.statistical-coaching.ch</u> (until 2015) | Probability & Statistics Regression & GLM Optimization Quantitative Methods Stochastic Processes (discrete & continue) Test theory Extreme values | Clinical Trial Management IS: Wikipedia, Sail, Comparis Earthquake Climate Finance Re-Insurance |
| 2009-2010 VUA-NL | | Realistic Neural Connectivity Networks model | | |
| 2008-2009 ITB- Indonesia | Mathematical stats Statistics for engineers & scientists | | | |
| 2003-2008 Leiden–NL ETHZ–CH | Computational statistics with R Mathematical stats Master-thesis | Classification Multivariate <u>discriminant</u> analysis Pattern recognition Machine Learning | | |
| 2000-2003 CWI-NL | | Probability of ruin model Bootstrap | | |
| 1995–2000 ITB- Indonesia | Basic Statistics Basic Probability Calculus | Winning research grant to NL Master in Math/Stat with scholarship | | |



Who is organizing this workshop?



Prof. Oscar Nierstrasz, University of Bern

Mascha Kurpicz-Briki, University of Bern





Prof. Pascal Felber, University of Neuchâtel

> Nevena Milojkovic, University of Bern



Statistical

Please introduce yourself too

- Your name
- A few words about your research project(s)

| First name | Last name | First name | Last name |
|-------------|-----------|------------|--------------|
| Antoine | Bellwald | Mirco | Kocher |
| Ionel Tudor | Calistru | Mascha | Kurpicz |
| Michele | Catasta | Alain | Mermoud |
| Claudio | Corrodi | Nevena | Milojkovic |
| Ana | De Abreu | Oscar | Nierstrasz |
| Ljiljana | Dolamic | Haidar | Osman |
| Mohammad | Ghafari | Iuliia | Proskurnia |
| Amit | Gupta | Jonnahtan | Saltarin |
| Yaroslav | Hayduk | Kasun | Samarasinghe |
| Aigul | Kaskina | Yuriy | Tymchuk |

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Do not hesitate...

- to interrupt me
- to participate actively
- to say

Bernadetta,

- You talk too fast...
- Can you please repeat this and that...
- I don't understand why...
- How did we get that results/numbers/values...
- I think you are wrong... (who knows \bigcirc)



2015, remember?

Statistical Inference: drawing conclusion about population from sample with some calculated uncertainty





What we have learned in 2015?

- 1. Data Analysis & Producing
- 2. Introduction to Probability
- 3. Foundation for Statistical Inference
 - Parameter vs Statistics
 - Sampling Distributions
 - Point Estimation
 - Set Estimation
 - Hypothesis Testing
- 4. Two-Way Table and One-Way ANOVA
- 5. Multiple Linear Regression
- 6. Binary Logistic Regression

And many R commands...



Data Analysis = Function Estimation

Data = numbers with context

Big Picture

- consists of variables and cases/observations
- data ≠ information



Y: dependent, response, outcome, output

X: independent, explanatory, predictor, input

We assume

$$Y = f(X) + \varepsilon$$

 ε is random part with $E[\varepsilon] = 0$ and $Var[\varepsilon] = \sigma^2$ f is a function connecting Y to X

But f is unknown, want to estimate it with \hat{f}



Function Estimation

Big Picture

- What is the goal/purpose of estimating *f*?
 - To describe how *Y* depends on
 - To explain causal effect of X_1, X_2, \dots, X_p on Y (testing causal theory)
 - Predicting Y from $X_1, X_2, ..., X_p$ (association)
- How to choose an estimate \hat{f} ?
- What criterion to assess the performance of an estimate \hat{f} ?



Data sets we have today

| | Data set 1 <i>critic</i> (from Yuriy) | Data set 2 <i>bug</i> (from Haidar) | Data set 3 <i>variabletype</i> (from Nevena) |
|---------|--|--|--|
| Goal | Explanation Hypotheses test | Prediction | Explanation Hypotheses test |
| Methods | Simple & Piecewise Linear Regression | GLM & Feature Selection | Analysis of Variance (ANOVA) - Two way with repeated measures |



Linear Regression Models

- Simple, easy to understand
- For explanation and description purpose, often provide an adequate and interpretability description on if and how the inputs affect the output
- For prediction purpose they can sometimes outperform fancier nonlinear models, especially in situations with:
 - Small numbers of training data
 - Low signal-to-noise ratio
 - Sparse data ($n \ll p$)
- Moreover, linear methods can be applied to transformations of the inputs and this expands their scope, e.g., basis function methods



General steps of data analysis

