

Complex Data Analysis

Syllabus Spring 2018

A. Instructor:

Jaroslav Harezlak, Associate Professor

Contact information:

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Office hours: By appointment

B. Combined lectures by Zoom (Skype) and in-class

February 19th, 2018 till June 15th, 2018

C. Course Description:

This course covers modern methods for the analysis of repeated measures, correlated outcomes and longitudinal data, including the unbalanced and incomplete data frequently encountered in biomedical research. Topics include an introduction to the analysis of correlated data, repeated measures analysis of variance (ANOVA), random-effects and growth-curve models, generalized linear models for correlated data, including generalized estimating equations (GEE), and generalized linear mixed models (GLMMs).

Class presentations and homework assignments will focus on data analysis in **R** (www.r-project.org) using functions *lme()* and *gls()* from the *library(nlme)*.

D. Prerequisites:

Linear regression course. Some familiarity with R statistical computing environment is assumed. Students who are uncertain about their level of preparation are encouraged to contact the instructor.

E. Required and Recommended Texts:

The course will use the material from following textbooks:

Fitzmaurice, G. M., Laird, N. M., and Ware, J. H. (2011) *Applied Longitudinal Analysis* Wiley & Sons (FLW), 2nd edition.

Laird, N.M. (2004) *Analysis of Longitudinal and Cluster-Correlated Data*, Institute of Mathematical Statistics

We will supplement the textbook with instructors' own lecture notes.

F. Educational objectives:

At the end of the course students will be able to:

- use modern statistical methods for the analysis of repeated measures, correlated outcomes and longitudinal data
- utilize R software for the analysis of such data

Format: Lectures with periodic labs to demonstrate and explore statistical methods and their implementation in R.

G. Evaluation and Grading:

Students will be evaluated based on their performance on the homework assignments (30%), the mid-term exam (30%), and the final project/presentation (40%).

H. Course Outline

Part I: ANALYSIS OF VARIANCE AND INTRODUCTION TO CORRELATED DATA

1. Lecture: Introduction
Reading: FLW, Chapter 1
2. Lecture: Introduction to Correlated Data
Reading: FLW, Chapter 2

Part II: LINEAR MODELS FOR LONGITUDINAL DATA

3. Lecture: Longitudinal Data – Linear Models
Reading: FLW, Chapter 3
4. **Lab**: Using functions `gls()` and `lme()` in R
- 5, 6. Lectures: Statistical Basis of Repeated Measures Analysis
Reading: FLW, Chapter 4
7. Lecture: Parallel Groups Repeated Measures Analysis
Reading: FLW, Chapter 5
8. **Lab**: Longitudinal Data Analysis using `gls()` + `lme()` in R
9. Lecture: Statistical Modeling of Repeated Measures
Reading: FLW, Chapter 6
10. Lecture: General Linear Model for Longitudinal Data
Reading: FLW, Chapter 6
11. Lecture: Mixed Effects Model for Longitudinal Data
Reading: FLW, Chapter 8

12. Lecture: Prediction of Subject-Specific Regression Coefficients
Growth Curve Models and Two-Stage Analysis
Reading: FLW, Chapter 8
13. **Lab**: Linear Mixed Models
14. Lecture: Selection of Model for Covariance, Empirical Variance Estimation,
Reading: FLW, Chapter 7
15. Lecture: Drop-outs, missing values and mis-timed values
Reading: FLW, Chapter 17
16. **Mid-Term Exam**
17. Lecture: Design and sample size estimation
Reading: FLW, Chapter 20

Part III: GENERALIZED LINEAR MODELS FOR LONGITUDINAL DATA

18. Lecture: Review of Logistic and Poisson Regression
19. Lecture: Introduction to Generalized Linear Models
Reading: FLW, Chapter 11
20. Lecture: Generalized Linear Models for Longitudinal Data
- 21-22. Lectures: Marginal Models and Generalized Estimating Equations
Reading: FLW, Chapters 12-13
23. **Lab**: GEE using R functions
24. Lecture: Generalized Linear Mixed Models
Reading: FLW, Chapter 14
25. Lecture: Contrasting Marginal and Mixed Effects Models
Reading: FLW, Chapter 16
26. **Lab**: GLMM using R function: gam() from the library(mgcv)
27. Lecture: Multilevel Models
Reading: FLW, Chapter 22
28. **Lab**: Multilevel Models
29. Student presentations