



Study Guide

Longitudinal and Correlated Data

Semester 2, 2026

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Longitudinal and Correlated Data (LCD)

Semester 2, 2026

Contact details of unit co-ordinator

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Background

Longitudinal and correlated data arise in many settings in health and medical research. Common examples include studies involving repeated measurements of individuals over time, in clinical trials and cohort studies, and cluster-randomised trials where participants are clustered within natural units such as schools or medical practices. The common characteristic of these data structures is that of correlated measurements either within an individual or within a cluster of individuals. Standard methods of statistical analysis assume independent observations and therefore do not accommodate this correlation, and more sophisticated methods need to be considered. There have been significant developments in these methods and their availability in statistical software packages in recent decades.

Context within the program

This unit builds on the knowledge and skills that students gained in the unit Regression Modelling for Biostatistics 1 (RM1), particularly linear regression for continuously-valued outcomes and logistic regression for binary outcomes. To accommodate the correlated data structures that typically result from longitudinal and cluster studies requires extending both the statistical models (that provide an idealised generative process for the data) and the estimation methods (that are applied to data to estimate the model parameters).

Prerequisites

Epidemiology (EPI), Mathematical Foundations for Biostatistics (MFB), Principles of Statistical Inference (PSI), Regression Modelling for Biostatistics 1 (RM1).

For University of Melbourne students the corresponding units are Epidemiology 1 (POPH90014 = EPI), Probability and Inference in Biostatistics (MAST90100 = PSI), Foundations of Regression (MAST90102 = RM1) and Advanced Regression (MAST90099 = RM2).

Co-requisites

Nil

Unit summary

This unit covers statistical models for longitudinal and correlated data in medical research. The concept of hierarchical data structures is developed, together with simple numerical and analytical demonstrations of the inadequacy of standard statistical methods. Beginning with models based on normal distributions, appropriate statistical methods involving generalised estimating equations and mixed linear models are developed and explored using the R and Stata statistical software packages. The limitations of traditional repeated measures analysis of variance are briefly discussed. Extensions to non-normal outcomes are developed and using a set of case studies, approaches based on generalised estimating equations (GEE) and generalised linear mixed models (GLMM) are developed and contrasted. Throughout, emphasis is placed on interpretation issues focussing on the underlying clinical or public health research question.

Workload requirements

The expected workload for this unit is 10-12 hours per week on average, consisting of guided readings, discussion posts, independent study, and completion of assessment tasks.

Learning Outcomes

At the completion of this unit students should be able to:

1. Recognise the existence of correlated or hierarchical data structures, and describe the limitations of standard methods in these settings
2. Develop and analytically describe appropriate models for longitudinal and correlated data based on subject matter considerations
3. Be proficient at using statistical software packages (Stata and R) to fit models and perform computations for longitudinal data analyses, and to correctly interpret results
4. Express the results of statistical analyses of longitudinal data in language suitable for communication to medical investigators or publication in biomedical or epidemiological journal articles

Unit content

The unit is divided into 6 modules, summarised in more detail below. Each module will involve 1, 2 or 3 weeks of study and includes the following material:

1. Module notes describing concepts and methods, and including some exercises of a more “theoretical” nature;
2. Selected readings from published articles or textbooks; and
3. One or more extended examples illustrating the concepts/methods introduced in the notes and including more practically oriented exercises.
4. An online ‘live’ tutorial, with date and time chosen by a Doodle poll of student preferences. The recording of the tutorial will be made available online for students unable to make it to the live session, or for later viewing.

Study materials for all Modules are accessed from the eLearning unit site. Assignments and supplementary material such as datasets will be available within each Assignment item. Please note that we are not able to post copies of copyright material (journal articles and book extracts)—for these you will have to rely on your home university’s library.

Recommended approaches to study

Students should work through each module systematically, following the module notes and any readings referred to, and working through the accompanying exercises. *You will learn a lot more efficiently if you tackle the exercises systematically as you work through the notes.* You should also work through all the computational examples in the notes for yourself on your own computer.

Outline solutions to the exercises in each module (except those to be submitted for assessment, as described below) will be posted online at the midway point of the allocated time for the module. This is intended to encourage you to attack the exercises independently (or via the eLearning site), and yet not make you wait too long to see the sketch solutions.

Make the most of this unit by engaging with coordinators and fellow students on the Canvas Discussion Board and in Tutorials. These are spaces to discuss the course material and related ideas and students are encouraged to make the most of them by engaging in respectful discussion. Furthermore, using Canvas for content-related communication and problem-solving will enable other students to benefit from responses and indeed to respond themselves, and we try to encourage as much interaction as possible within the class through this medium

Questions about Assignments should be directed to the coordinators in the first instance to avoid any Academic Honesty issues.

Method of communication with coordinator(s)

Questions about administrative aspects or course content can be emailed to the coordinators. Please use “LCD:” in the Subject line of your email to assist in keeping track of our email messages. Coordinators will be available to answer questions

related to the module notes and practical exercises, and to address any other issues that require clarification.

Please note that instructors are not necessarily available every day of the week and you should expect that it may take a day or so to respond to questions (possibly longer over weekends and during breaks).

Module descriptions

Below is an outline of the study modules, followed by a timetable and assessment description table

Each module is scheduled to begin on a Monday and conclude on the Sunday of the following week.

Module 1: Introduction: paired data and simple clustering

- recognise situations where there is clustering in a data structure
- appreciate the impact that clustering can have on standard analytical methods for uncorrelated data
- recognise when research questions involve between-cluster or within-cluster comparisons
- formulate an appropriate regression model, incorporating random effects, for paired data and understand the interpretation of parameters in this model
- understand the principles of generalised estimating equations (GEE) and be able to manually perform its “behind-the-scenes” calculations in a simple model for the mean of a continuous outcome variable when there is within-cluster correlation but no covariates
- gain an introduction to basic commands in Stata and/or R for performing correlated data analyses.

Module 2: Exploratory analysis and simple methods using summary measures

- use graphical methods to explore patterns in longitudinal data
- compute simple summary statistics to describe patterns in longitudinal data
- convert longitudinal data from “long” to “wide” format and vice versa
- understand the distinction between longitudinal and cross-sectional relationships
- appreciate the impact that correlated data has on sample size calculations in the case of cluster randomised trials

Module 3: Modelling longitudinal continuous outcomes: estimating equations (marginal model) approach

- understand how conventional multiple linear regression (ordinary least squares) can be extended to model the mean of a continuous outcome variable while allowing for correlated observations within clusters
- appreciate how such models can be specified simply via matrix representations
- understand the form of simple correlation structures and recognise situations when they are likely to be appropriate
- understand the general concept, construction and representation of weighted least squares estimators as solutions to “estimating equations”
- understand how residuals are used to estimate within-cluster correlation parameters and the general method of solution of estimating equations
- recognise the usefulness and understand the construction of robust variance estimators

Module 4: Modelling longitudinal continuous outcomes: mixed models approach.

- understand how random effects can be incorporated into regression models to represent certain correlation structures within each cluster
- appreciate the difference between these likelihood-based mixed-model methods and those of marginal models for the mean (as in Module 3)
- be proficient in the use of Stata and/or R commands to estimate parameters of mixed models with random intercepts and slopes and a variety of correlation structures
- be able to apply criteria to assist in selection of an appropriate correlation structure

Module 5: Methods for discrete data: GEE and generalized linear mixed models (GLMM)

- appreciate how the generalised estimating equation framework is an extension of logistic regression to accommodate within-cluster correlation with binary data
- appreciate how likelihood-based inferences are possible using generalised linear mixed models incorporating random effects, and of general issues involved in their computation
- understand the difference in interpretation of regression parameters between GEE and generalised linear mixed modelling approaches
- be proficient in the fitting of both sorts of models in Stata and/or R

Module 6: Methods for longitudinal & correlated count data

- appreciate how the generalised estimating equation framework can accommodate count data with within-cluster correlation
- understand the phenomenon of overdispersion with count data and how it can be accommodated in estimation methods

Unit schedule

Semester 2, 2026 starts on Monday 27th July 2026

Week	Week commencing	Module	Topic	Assessment
1	July 27	1	Introduction: paired data and simple clustering	
2	Aug 3	1		
3	Aug 10	2	Exploratory analysis and simple methods using summary measures	
4	Aug 17	2		Modules 1 – 2 assessment due 11:59pm August 24
5	Aug 24	3	Modelling longitudinal continuous outcomes: estimating equations (marginal model) approach	
6	Aug 31	3		
7	Sep 7	3		Modules 1 – 3 assessment due 11:59pm September 14
8	Sep 14	4	Modelling longitudinal continuous outcomes: mixed models approach	
	Sep 21	4		
9	Sep 28		Mid Semester break	
10	Oct 5	5	Methods for discrete data: GEE and generalized linear mixed models (GLMM)	Module 4 assessment due 11:59pm Tues October 6
11	Oct 12	5		
12	Oct 19	6	Methods for longitudinal & correlated count data	Modules 4 – 6 assessment due 11:59pm Friday November 6

Assessment

- Two sets of modules exercises worth 20% each, to be made available after Module 2 and Module 4, and to be completed within approximately two weeks; and
- Two written reports worth 30% each, to be made available after Module 3 and Module 6, and to be completed within approximately three weeks.

Assessments are due by 11:59pm on the stated day.

Assessment name	Assessment type	Coverage	Learning objectives	Weight
Module 1-2 assessment	Short-answer questions	Module 1-2	1-2	20%
Module 1-3 assignment	Report	Module 1-3	1-3	30%
Module 4 assessment	Short-answer questions	Module 4	2-4	20%
Module 4-6 assignment	Report	Module 4-6	1-4	30%

In general, you are required to submit your work as a document prepared in Microsoft Word or something similar. We strongly recommend you become familiar with equation typesetting software such as Microsoft's Equation Editor (or use LaTeX) for algebraic work. You may submit neatly handwritten work, however, please note that marks will potentially be lost if the solution cannot be understood by the markers due to unclear or illegible writing. Handwritten work should be scanned and collated into a single pdf file and submitted via the eLearning site. See the [BCA Assessment Guide](#) for guidelines on acceptable standards for assessable work.

Students are encouraged to discuss relevant topics in the Discussion Board. Please avoid posting questions relating directly to assessable material. These should be emailed to the Unit Coordinator in the first instance.

Explicit solutions to assessable exercises should not be posted for others to use. Each student's submitted work must be clearly their own, with anything derived from other students' discussion contributions clearly attributed to the source.

Submission and academic honesty policy

All assessment material should be submitted via the relevant Assessment module in Canvas unless otherwise advised. Turnitin plagiarism detection is applied to all submissions. For detailed information, please see the [BCA Assessment Guide](#), which includes links to the Academic Honesty policies at member universities. Please familiarise yourself with the procedures and policies at your home university. You will need to indicate your compliance with the plagiarism guidelines and policy at your home university.

Use of ChatGPT, Claude or other generative AI tools

The assessment tasks in this Unit have been designed to be challenging, authentic and complex. Although individual assessment components may provide specific guidance regarding the use of generative artificial intelligence (AI) tools (e.g., ChatGPT), successful completion of these components will require students to critically engage in specific contexts and tasks for which AI will provide only limited support and guidance. In all cases, a failure to reference the use of generative AI may constitute student misconduct under the Student Code of Conduct of your University of enrolment. To successfully complete assessment tasks, students will be required to demonstrate detailed comprehension of their written submission independent of AI tools.

Late submission and extension procedure

The standard BCA policy for late penalties for submitted work is a 5% deduction from the available mark for each day the assessment is late, up to a maximum of 10 days (including weekends and public holidays). Extensions are possible, but these need to be applied for (by email) as early as possible. The Unit Coordinator can approve extensions up to three days; for extensions beyond three days, you must apply to your home university, using their standard procedures.

Learning resources

There is no single prescribed text for the subject, but a few reference books are recommended as background material (list below). The first book in the list is the one that we find closest to our approach in LCD (although it appeared after the first draft of the course was written), so if you were to obtain one book this would be our recommendation. The module notes and case studies form the primary material for this subject, and required readings from selected texts, are provided.

Fitzmaurice G, Laird N, Ware J. *Applied Longitudinal Analysis*, John Wiley and Sons, 2004. [Note that a 2nd edition appeared in 2012. All readings for this semester are taken from the 2004 edition – they differ minimally from the 2012 edition]

Diggle PJ, Heagerty P, Liang K-Y, Zeger SL. *Analysis of Longitudinal Data*, 2nd Edition, Oxford University Press, 2002.

Singer JD, Willett JB. *Applied Longitudinal Data Analysis: Modeling Change and Event Occurrence*, Oxford University Press, 2003.

Verbeke G, Molenberghs G. *Linear Mixed Models for Longitudinal Data*, Springer, 2000.

Brown H, Prescott R. *Applied Mixed Models in Medicine*, 3rd Edition, Wiley, 2015.

Software requirements and assumed knowledge

For this subject you will need to have access to, and a working familiarity with, either Stata or R.

Students using Stata will need at least version 16 that was released in July 2019. The current version is Stata 19 released in April, 2025. We are not aware of any major differences between Stata versions that affect the material, but minor issues will be pointed out in *Canvas* postings. Importantly, whichever version you are using, please ensure that you have performed the online update to the latest update of that version. (Use the command `update query`).

For R, the notes assume you are working with the latest version, although slightly earlier versions should not have any important differences. For help with R, please see [Learning R](#) in the Student Resources site.

Required mathematical background

No additional mathematical background is required beyond what is covered in Mathematical Foundations for Biostatistics (MFB) [including its matrix algebra section], Principles of Statistical Inference (PSI), Regression Modelling for Biostatistics 1 (RM1).

Feedback

Our feedback to you

The types of feedback you can expect to receive in this unit are:

- Formal individual feedback on submitted exercises assignments
- Discussions during Module tutorials
- Responses to questions posted on *Canvas*

Your feedback to us

One of the formal ways students provide feedback on teaching and their learning experience is through the BCA student evaluation survey at the end of each semester. The feedback is anonymous and provides the BCA with evidence of aspects that students are satisfied with and areas for improvement.

Unit changes, including response to recent student evaluation

LCD was last delivered in Semester 1 2026. Apart from new assessment tasks, there have been only minor changes since that delivery in the form of correcting typographical errors and minor edits for greater clarification of the text.

Acknowledgments

Professor Andrew Forbes (Monash University) and Professor John Carlin (Murdoch Children's Research Institute (MCRI) and University of Melbourne) were jointly responsible for the development of the material for this unit. Dr John Holmes (University of Canterbury, New Zealand) added the data analysis examples using *R* in 2022.