

1 Scoping Review Protocol: Statistical Models for Longitudinal Data
2 in Health and Biomedical Research: Current State, Challenges,
3 and Opportunities

4 Ariel I. Mundo Ortiz

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35 1 Notes

36 As of Sept 7, 2022 this document follows the structure recommended by PRISMA-P
37 <https://prisma-statement.org/documents/PRISMA-P-checklist.pdf>

38 Scoping review is exploratory, can be a little broad but is best to start with one to make sure
39 that the method works, and that its not too biased because of dispairing standards within
40 subfields. Oncology, neurodevelopment, mental health: psichology, psychyatry

- 41 • Oncology can be the sandbox.

42 2 Registration

43 This section will be populated with the registration number and registry name once the protocol is submitted
44 for peer review.

45 3 Author Contributions

- 46 • AM: Writing, query design, data extraction and analysis . . .

47 Other authors to add later

48 4 Amendments

49 Protocol amendments resulting from peer review will be indicated in this section indicating the date of each
50 amendment.

51 5 Support

52 This section will indicate the sources of financial or other support for the review

53 5.1 Sources

54 6 Introduction

55 6.1 Rationale

56 Longitudinal studies are frequently used in the health sciences (biomedical research, epidemiology, public
57 health, among others) to examine the temporal effect of a treatment or intervention **add about those**
58 **studies where there is not an intervention, but follow up/evolution**^{1,2}. However, the statistical
59 analysis of longitudinal data requires to take into consideration factors such as data missingness, correlation,
60 and non-linear trends^{3,4}, which represent an “analytic cost” associated with the complexity of longitudinal
61 data².

One of the problems derived from the “analytic cost” of longitudinal data pertains the misspecification of the statistical models used to analyze such data (i.e., the use of models that are not coherent with the data), a problem that has been shown to occur frequently in the health sciences⁵. This problem with model misspecification can be linked to a historical preference by researchers to use the repeated measures analysis of variance (rm-ANOVA) as the default method to analyze longitudinal data, despite the fact that the multiple assumptions required by this model are frequently not satisfied by the data collected in longitudinal studies⁴.

On the other hand, multiple modern statistical models were developed during the last 30 years to address the limitations of rm-ANOVA. Linear mixed models, generalized additive mixed models, and generalized estimating equations are among these modern statistical models developed for longitudinal data^{6–10}. However, the use of such modern statistical methods has been the exception rather than the norm in the health sciences¹¹, even on this day and age where these modern methods have been brought to a wider audience with the development of computational tools such as Python or R.

Unfortunately, the misuse and lack of reproducibility of statistical analyses continue to be major problems in the health sciences^{12–15}. In the case of longitudinal data, where modern methods exist beyond rm-ANOVA that can help researchers obtain better inference from their data, there is a need to understand what are the trends in the adoption of these statistical methods in the health sciences to measure the adoption of reproducibility practices by the field at large, while also identifying the reasons that may cause researchers use avoid the use of modern statistical methods for longitudinal data.

7 Objectives

This study aims to:

- Identify the different statistical models for longitudinal data that are used in the health sciences in order to measure the current extent in the adoption of modern statistical methods by the field (Aim 1a)
- Summarize the computational tools used by researchers in the health sciences to statistically analyze longitudinal data to understand the current status of the field with regards to reproducibility. (Aim 1b)
- List statistical methods for longitudinal data developed within the last decade in order to showcase

newer methods that may be applicable for longitudinal data in a biomedical/health context. (Aim 2)
maybe a different database different from Web of Science? Database for Stats or Math?

8 Review Question

- What are the statistical methods used in biomedical/health sciences research?
- Has the use of modern statistical methods increased in the field during the last 20 years?
- What computational tools are most commonly used by researchers to analyze longitudinal data, and how in turn this affects reproducibility?
- What are most recent statistical methods developed for longitudinal data, and how can they be applied in the health sciences?

9 Methods

9.1 Types of Studies

For all the study aims, studies included in the analysis correspond to peer-reviewed publications in English.

9.2 Eligibility Criteria

9.2.1 For the Application of Modern Statistical Models on Longitudinal Biomedical/Health Data (Aims 1a and 1b)

9.2.1.1 Inclusion Criteria

Articles that are written in English, belong to the biomedical/health sciences fields, describe the collection and analysis of continuous or discrete longitudinal data, indicate the statistical model used to analyze the data, and report the results of their statistical analyses.

9.2.1.2 Exclusion Criteria

Cross-sectional studies, tutorials that present the application of existing statistical methods to biomedical/health data, reviews, meta-analyses, or systematic reviews on existing statistical methods for longitudinal data, studies that use only descriptive statistics to summarize/analyze the data, studies that collect

113 and analyze categorical data. **You don't want to exclude things right away, much rather get them**
114 **and then decide.**

115 **9.2.2 For Methods on Longitudinal Data (Aim 2)**

116 **9.2.2.1 Inclusion Criteria**

- 117 • Articles that:

118 Are written in English, present new methodologies or significant improvements to existing methods for
119 longitudinal data.

120 **9.2.2.2 Exclusion Criteria**

121 Systematic reviews, meta-analyses, or reviews of statistical methods for longitudinal data, tutorials that
122 present the application of existing statistical methods to biomedical/health longitudinal data.

123 **9.3 Information Sources**

124 Studies will be retrieved from PubMed and Web of Science.

125 **9.4 Search Strategy**

126 PubMed and Web of Science databases will be used. Below the full search strategy for PubMed is presented
127 for all the aims of the scoping review.

128 **9.4.1 For the Application of Modern Models on Longitudinal Biomedical/Health Data**

129 **9.4.1.1 PubMed**

130 **9.4.1.1.1 Query:**

131 (biomedical OR health) AND ((repeated measures) OR (longitudinal study) OR (longitudinal data))
132 AND ((statistical analyses) OR (statistical analysis)) NOT (Review[Publication Type] OR Meta
133 analy*[Publication Type]) NOT ("Statistics as Topic/methods"[Majr] OR "Statistics as Topic/statistics
134 and numerical data"[Majr] OR "Models, Statistical"[Mesh] OR "Research Design"[Mesh])

135 Hits: 10,972

136 9.4.2 For Methods on Longitudinal Data

137 9.4.2.1 PubMed

138 9.4.2.1.1 Query 1:

139 (“Models, Statistical” [Mesh] OR “Biostatistics/methods”[Mesh]) AND (“Longitudinal Studies”[Mesh])
140 NOT (Review[Publication Type] OR Meta Analys*[Publication Type] OR “editorial”[Publication Type])
141 NOT (“survival”[Title/abstract]) NOT (“tutorial”[title/abstract] OR “orientation”[title/abstract]) NOT
142 (Humans[Mesh] OR Adolescent [Mesh] OR Animals[Mesh])

143 Hits: 142

144 9.5 Data Collection and Analysis

145 9.5.1 Selection Process and Data Management

146 Two reviewers will independently analyze the database search results and pre-screen articles based on ti-
147 tle and abstract content following the aforementioned inclusion/exclusion criteria. Manuscripts from the
148 database(s) search will be stored in the Covidence platform, where duplicated entries will be removed. For
149 articles where pre-screening inclusion (or exclusion) is unclear based on title and abstract analysis, full-text
150 review will be used to make a decision following review by a third independent reviewer. Manuscripts
151 included after title and abstract pre-screening will be further screening by two reviewers that will indepen-
152 dently examine the full text of each article.

153 9.5.2 Data Collection Process

154 Pilot forms (electronic spreadsheets) will be tested using a representative sample of the studies to be
155 reviewed (~100 studies). Information in the forms will be independently included by each reviewer. The
156 forms will be updated (if needed), after the pilot test by consensus between the reviewers.

157 Information obtained from each study (statistical method used, software, etc.) will be tabulated indepen-
158 dently by the reviewers in an electronic spreadsheet.

159 **9.6 Data Items**

160 Aims 1a and 1b:

161 Statistical method used, sub-area of application (oncology, psychology, public health, etc), computational
162 tool used, congruence between statistical method used and the data, year of publication

163 Aim 2:

164 Statistical method reported, assumptions of the model, computational tools available for its implementation,
165 year of publication

166 **9.7 Risk of Bias in Individual Studies**

167 N/A

168 **9.8 Data Synthesis**

169 The data from the results of each included study will be extracted into electronic spreadsheets. Summary
170 measures for Aims 1a and 1b include plots (pie, bar, etc.) to show the relative use of each statistical
171 method reported, computational tool, and congruence between statistical method and the data. Each plot
172 will be segmented by year to show trends over time. Table 1 presents the headers of the pilot electronic
173 spreadsheet.

174 The pilot electronic spreadsheet can be found in the following link: [Pilot Spreadsheet](#)

175 For Aim 2, a table will be created where statistical method, year of publication, assumptions of the model,
176 and applicability to health data is reported.

177 **9.9 Meta-Biases**

178 N/A

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209 *Stud*. 2015;52(1):5-9.

Table 1: Pilot spreadsheet for data extraction

DOI	Title	Subfield	Journal	Question	Country	Source of Result (Data)	Year	Statistical Method	Software	Model assumptions checked?	Data/Model Congruency?	Code available?	Notes
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