

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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1 Notes

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

38 **2 Registration**

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

41 **3 Author Contributions**

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

43 Other authors to add later

44 **4 Amendments**

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

47 **5 Support**

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

49 **5.1 Sources**

50 **6 Introduction**

51 **6.1 Rationale**

52 Longitudinal studies are frequently used in the health sciences (biomedical research, epidemiology, public
53 health, among others) as they allow to examine how the temporal effect of a treatment or an intervention,
54 in contrast to a cross-sectional study, which only allows to examine the effect of the intervention at a single
55 time point. When compared their cross-sectional counterparts, longitudinal studies allow for increased
56 statistical power and more cost efficient strategies^{1,2}. However, the statistical analysis of longitudinal data
57 requires to take into consideration factors such as data missingness, correlation, and non-linear trends,

58 which do not occur on cross-sectional data^{3,4}. In other words, there is an “analytic cost” associated with
59 the increased complexity of longitudinal data².

60 This additional layer of complexity has led to a problem of model misspecification in the statistical analysis
61 of the data (i.e., the use of a statistical model that is not coherent with the data), which has been reported to
62 occur in many fields, including the health sciences⁵. For example, in a landmark study Liu et al. showed that
63 in a subset of papers in the biomedical sciences, the most popular model used to analyze longitudinal data
64 was the analysis of variance (ANOVA, an approach that fails to take into account the correlation between
65 measures over time), and that only 18% of the studies analyzed used models intended for longitudinal
66 analysis while checking that the assumptions of the model were satisfied by the data⁶.

67 Historically, the repeated measures ANOVA (rm-ANOVA, a statistical model for longitudinal data) has
68 been the preferred method in the health sciences to analyze longitudinal data, despite the fact that the
69 multiple assumptions required by this model are frequently not satisfied by the data collected in longitudinal
70 studies⁴. On the other hand, the last 30 years have seen incredible progress in the field of Statistics with the
71 development of statistical models for longitudinal data that relax the assumptions of rm-ANOVA. Linear
72 mixed models, generalized additive mixed models, and generalized estimating equations are among these
73 modern statistical models developed for longitudinal data^{7–11}. From these statistical methods, linear mixed
74 models and generalized estimating equations are the two classes of models that have been frequently applied
75 to analyze longitudinal data in the health sciences during the last decade^{12–14}.

76 However, modern statistical methods that are suited to analyze longitudinal data have been the exception
77 rather than the norm in the health sciences. In 2001, a study reported that only 30% of the clinical trials
78 analyzed used linear mixed models to analyze their results, and that the preferred method of analysis
79 continued to be rm-ANOVA¹⁵ (in comparison, McCullagh and Nelder’s seminal book on the generalized
80 linear model (GLM) was published in 1989¹⁶, and there was ongoing work on the extension of the GLM
81 framework to the mixed model case by 1993¹⁷). Apart from the aforementioned study, there are not recent
82 papers that examine the use of modern statistical methods for longitudinal data in the health sciences.
83 Such information is critical to understand if the use of these methods has increased or decreased in the field
84 over the last 20 years, and the reasons behind such changes.

85 Additionally, the reproducibility crisis is an ongoing issue in the health sciences^{18,19}, a major component of
86 it being the misuse and lack of reproducibility of statistical analyses^{20,21}. Despite the fact that the landscape
87 of statistical software has vastly increased in the last decade with many statistical computational tools now
88 available to researchers, reproducibility standards vary between each computational tool²². Furthermore,

there is still high variability in the amount of statistical reporting across journals²³. Understanding what statistical computational tools are used nowadays by researchers in the health sciences can provide an assessment of the advances in the field towards research reproducibility, while identifying limitations that might still be in place.

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)

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?

110 **9 Methods**

111 **9.1 Types of Studies**

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

113 **9.2 Eligibility Criteria**

114 **9.2.1 For the Application of Modern Statistical Models on Longitudinal Biomedical/Health** 115 **Data (Aims 1a and 1b)**

116 **9.2.1.1 Inclusion Criteria**

- 117 • Articles that:
 - 118 – Are written in English
 - 119 – Belong to the biomedical/health sciences fields
 - 120 – Describe the collection and analysis of continuous or discrete longitudinal data
 - 121 – Indicate the statistical model used to analyze the data
 - 122 – Report the results of their statistical analyses

123 **9.2.1.2 Exclusion Criteria**

- 124 • Cross-sectional studies
- 125 • Tutorials that present the application of existing statistical methods to biomedical/health data
- 126 • Reviews, meta-analyses, or systematic reviews on existing statistical methods for longitudinal data
- 127 • Studies that use only descriptive statistics to summarize/analyze the data
- 128 • Studies that collect and analyze categorical data

129 **9.2.2 For Methods on Longitudinal Data (Aim 2)**

130 **9.2.2.1 Inclusion Criteria**

- Articles that:

- Are written in English

- Present new methodologies or significant improvements to existing methods for longitudinal data

9.2.2.2 Exclusion Criteria

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

9.3 Information Sources

Studies will be retrieved from PubMed and Web of Science.

9.4 Search Strategy

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

9.4.1.1 PubMed

Note: I removed the query I had with specific names of specific statistical models (as the results were too broad and non-specific).

9.4.1.1.1 Query 2:

(biomedical OR health) AND ((repeated measures) OR (longitudinal study) OR longitudinal data) AND ((statistical analyses) OR (statistical analysis)) NOT ~~((review) OR (meta-analysis))~~ when you put NOT that might exclude papers with Review and meta analysis as words in the paper

Hits: 12,617

Response to comment: I followed your advice and re-wrote the query, but now I was sure to exclude meta-analysis and review papers by type of publication, and papers that are classified in PubMed as devoted to Statistical methodologies (not about application of methods to longitudinal data):

153 **9.4.1.1.2 Query 3 (modified Query 2):**

154 biomedical OR health) AND ((repeated measures) OR (longitudinal study) OR longitudinal data)
155 AND ((statistical analyses) OR (statistical analysis)) NOT (Review[Publication Type] OR Meta
156 analy*[Publication Type]) NOT (“Statistics as Topic/methods”[Majr] OR “Statistics as Topic/statistics
157 and numerical data”[Majr] OR “Models, Statistical”[Mesh] OR “Research Design”[Mesh])

158 Hits: 10,948

159 Comments: [This query is better than Query 2.](#)

160 Papers from this query appear to be good. The query catches many papers from psychology and psychiatry,
161 but the ones I checked did said used linear mixed models or regression in their analyses. A few of them still
162 deal with methodologies, but seems to be much more less than in the previous query.

163 **9.4.1.2 Web of Science**

164 **9.4.1.2.1 Query 1:**

165 WC=(biom* OR health OR allergy OR cell biology OR cardio* OR hematology OR immunology OR life
166 sciences biomedicine other topics OR medical informatics OR neuro* OR oncology OR pharmacology OR
167 radiology, nuclear medicine & medical imaging OR research & experimental medicine OR substance abuse
168 OR optics) AND AK=(longitudinal study OR repeated measures study) NOT ALL=(review OR meta
169 analysis) NOT AK=(model* AND study design) NOT KP=(model)

170 Hits: 4,716

171 [when you put NOT that might exclude papers with Review and meta analysis as words in the paper](#)

172 [Remove the NOT and check the suggestions for the PubMed section](#)

173 **9.4.1.2.2 Query 2 (Updated Query 1):**

174 WC=(biom* OR health OR allergy OR cell biology OR cardio* OR hematology OR immunology OR life
175 sciences biomedicine other topics OR medical informatics OR neuro* OR oncology OR pharmacology OR
176 radiology, nuclear medicine & medical imaging OR research & experimental medicine OR substance abuse
177 OR optics) AND AK=(longitudinal study OR repeated measures study) NOT AK=(model* AND study
178 design) NOT KP=(model) NOT TI=(review OR meta analy*)

179 Comments: [Updated this query based on your comments.](#)

180 Hits: 4,595

181 I updated this query based on your comments. Now I used NOT but focused on the title (hence, TI) so
182 articles with “review” or “meta analysis” are excluded. An additional filter is that Web of Science allows
183 to select the type of publications so I chose “Article”, and this would remove review papers, editorials or
184 other materials that are not research papers.

185 9.4.2 For Methods on Longitudinal Data

186 you need to describe this part in your objectives

187 I re-wrote the query here based on your comments. After reading your comments and the queries
188 I created, I realized that the filtering was not correct. I wrote a new query that I believe better represents
189 the terms we want to look at.

190 9.4.2.0.1 Query 1:

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

195 Hits: 142

196 Comments:

197 The rationale for this query is to find papers that have been labeled as dealing with models in Biostatistics
198 or Statistics, that deal with longitudinal data, but excluding reviews, editorials, meta analysis, tutorials
199 (that show how to implement an existing model, but not the development of a new model). Additionally,
200 I added the “humans”, “adolescent”, and “animal” labels to exclude, because there are **many** papers that
201 have all the previous labels but that are devoted to comparing methods, or about studies with animal or
202 clinical data (without those last filters for humans, adolescents, and animals the hits are 14,702).

203 Again, papers that describe the development of new methods for longitudinal data should be relatively few
204 when compared to papers that deal with application, and that is why to me the result of the query (142
205 hits) makes sense. I did take a look at the papers of this query and all of them seem to be about models,
206 which is what we want.

207 9.4.2.1 Web of Science

208 9.4.2.1.1 Query 1:

209 AK=((longitudinal OR repeated measures **soutOR longitudinal data, already included in longitudinal**) AND
210 (model OR design OR **method**)) ~~NOT ALL=(review OR meta analysis) NOT ALL=(survival analysis)~~

211 Hits: 3,071

212 Comments: [This query seems to be good.](#)

213 This query returns papers that deal with methods for longitudinal analysis. Two additional options can be
214 selected: 1) include only articles (which reduces the number of hits to 2,936 as book chapters and editorials
215 are omitted) and 2) select from the 01/01/2000 until today (which could be reasonable as the increment of
216 models has occurred during the last two decades. This option reduces the number to papers to 2,849).

217 **we will see when the queries is revised what is the best strategy**

218 9.4.2.1.2 Query 2 (updated Query 1):

219 I updated the query based on your comments. The query is more specific now, please see below.

220 AK=((longitudinal OR repeated measures) AND (model OR design OR method)) NOT TI=(review OR
221 meta analysis) NOT AK=(survival analysis) AND SU=(stat* OR MATH*) NOT AB=(tutorial)

222 Hits: 1,978

223 Comments:

224 This query looks for papers where the authors used the keywords longitudinal or repeated measures, and
225 model, or design or method; excludes papers that have “review” or “meta analysis” in the title, excludes
226 those that have keywords of “survival analysis”, and searches papers where the main subject has been
227 tagged as statistics or math. Finally, papers that have the word “tutorial” in the abstract are excluded as
228 well.

229 9.5 Data Collection and Analysis

230 9.5.1 Selection Process and Data Management

231 Two reviewers will independently analyze the database search results and pre-screen articles based on ti-
232 tle and abstract content following the aforementioned inclusion/exclusion criteria. Manuscripts from the

233 database(s) search will be stored in the Covidence platform, where duplicated entries will be removed. For
234 articles where pre-screening inclusion (or exclusion) is unclear based on title and abstract analysis, full-text
235 review will be used to make a decision following review by a third independent reviewer. Manuscripts
236 included after title and abstract pre-screening will be further screening by two reviewers that will indepen-
237 dently examine the full text of each article.

238 **9.5.2 Data Collection Process**

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

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

244 **9.6 Data Items**

245 Aims 1a and 1b:

- 246 • Statistical method used
- 247 • Sub-area of application (oncology, psychology, public health, etc)
- 248 • Computational tool used
- 249 • Congruence between statistical method used and the data
- 250 • Year of publication

251 Aim 2:

- 252 • Statistical method reported
- 253 • Assumptions of the model
- 254 • Computational tools available for its implementation
- 255 • Year of publication

256 9.7 Risk of Bias in Individual Studies

257 N/A

258 9.8 Data Synthesis

259 The data from the results of each included study will be extracted into electronic spreadsheets. Summary
260 measures for Aims 1a and 1b include plots (pie, bar, etc) to show the relative use of each statistical method
261 reported, computational tool, and congruence between statistical method and the data. Each plot will be
262 segmented by year to show trends over time.

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

265 9.9 Meta-Biases

266 N/A

267 10 References

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269 [30:4%3C330::AID-PPUL10%3E3.0.CO;2-D](https://doi.org/10.1002/1099-0496(200010)30:4%3C330::AID-PPUL10%3E3.0.CO;2-D)
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