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Data Analysis with Competing Risks and Intermediate States



Ronald B. Geskus



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Chapman & Hall/CRC Biostatistics Series

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To life



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Preface

In the end we all die. More interesting than the death event itself is the time component: at what age does one die and what characteristics make some individuals die earlier than others? Survival analysis provides the set of tools that help answer the question of which factors influence the time until the occurrence of some event, which is not restricted to be death.

In the end we all die, but not all from the same cause. Information on the spectrum of causes of death has added value. Figure 1 shows the number of individuals that died of different causes in the twentieth century. The smallest



FIGURE 1

All causes of death in the 20th century.

subgroups may not be readable (have a look at the website if you want to read them¹), but it is seen that the most frequent category is made up of the noncommunicable diseases, within which cardiovascular diseases form the largest subgroup. The causes of death are competing: if one occurs, the others do not. Again, including the time component has added value. Some causes, like measles, tend to occur at a young age, whereas others, like prostate cancer, almost exclusively occur in older men. Measles mortality prevents prostate cancer from occurring, but if mortality due to measles is reduced, mortality due to prostate cancer, at a later age, may rise. A competing risks model extends the classical survival setting by considering a collection of mutually

¹The figure was downloaded on September 1st, 2013 from the website http://www.informationisbeautiful.net/visualizations/20th-century-death. Author: David McCandless.

exclusive potential event types. Different causes of death are the classical example, but any set of event types can be considered.

The issues that come up when competing risks are present have often been ignored, even in top (medical) journals [61]. But the situation is changing rapidly. In the last decade, several papers have been published that explain when and why a standard time-to-event approach fails to provide the correct answer. Still, confusion remains with respect to the quantities that can be estimated and their interpretation. Fortunately, once the concepts are understood and the appropriate type of analysis has been chosen, techniques of estimation are not much different from the ones used in classical survival analysis with only one event type.

In the end we all die, but not all for the same reason nor with the same life histories. Even if two individuals die at the same age and of the same cause, their life courses have been different. Between birth and death, intermediate events occur that influence one's life course. The are several ways to model the occurrence of such events and their effect on later ones. One approach is via a multi-state model, in which events are seen as transitions from one state to another. Death may be the final one, but the process under investigation may terminate earlier. Under the frequently assumed Markov property, a multistate model can be seen as a sequence of competing risks models.

Multi-state models can give a more detailed description of a disease process. Only few articles have explained the use of multi-state models to nonstatisticians. Although the range of modeling choices and possible assumptions is larger than in a competing risks setting, there is little additional complexity with respect to interpretation. Also, some of the estimation techniques are a direct extension of those from the competing risks setting such that the same software can be used. Other computations are more complex and require software that has been written specifically for such models.

This book is divided into five chapters and an epilogue. The first chapter introduces the main concepts, with emphasis on the competing risks setting. For a number of examples that will be used throughout the book, we formulate the type of questions that may be of interest and define the corresponding estimands. We explain when the data at hand can be used to answer these questions. We also give an overview of the main definitions and techniques from classical survival analysis that are used and extended in later chapters. In Chapters 2 and 3 we more formally define the concepts that play a role in competing risks and multi-state models respectively. We address nonparametric estimation of the relevant quantities. In Chapters 4 and 5, we quantify the effect of covariables via regression models. In Chapter 4 we explain how and why the ideas and techniques of the classical Cox proportional hazards model extend to settings with competing risks and intermediate states. The difference is in the interpretation of the estimates and their translation to the cumulative scale. The latter is explained in Chapter 5. We also describe two approaches in which parameter estimates have a direct interpretation on the cumulative scale. One quantifies effects on another type of hazard, the sub-

Preface

distribution hazard. This model is often called the Fine and Gray model. The third uses a model that quantifies effects directly on the cumulative probability, the proportional odds model. Two of the main difficulties from a practical perspective are the translation of research questions into modeling choices and the interpretation of the results. Therefore, these issues are given a lot of attention throughout the book. The epilogue is completely devoted to this issue.

The last four sections of each chapter have the same structure. We first summarize the concepts that have been introduced, place them in a broader perspective and refer to issues that come up in subsequent chapters. The next section contains exercises that are intended to help understand and reflect on the concepts. Technical exercises are denoted by an asterisk. Answers to the exercises are provided at the end of the book. The exercises are followed by a section devoted to software. We briefly explain options in SAS and Stata. We give a detailed description of the functionality in the R statistical program [83]. Each chapter ends with computer practicals, in which you are asked to practice the concepts in R, using an existing data set on patients that underwent a bone marrow transplantation. In principle, all of the practicals on competing risks and some on multi-state models can also be made using Stata or SAS, but we do not provide any suggestions or answers.

Since we use examples from the biomedical and epidemiological field, the intended readership primarily consists of medical statisticians and epidemiologists. However, the book is useful for any researcher that has some experience with the analysis of standard time-to-event data and wants to extend knowledge and skills to the competing risks or multi-state setting. It should be easy to translate "individuals" and "diseases" to units and phenomena from one's own research area.

With respect to the topics covered and the intended readership, our book is most closely related to the book *Competing Risks and Multistate Models* with R [12]. That book takes a more theoretical perspective and relies fairly heavily on the description via counting processes, whereas our focus is more on interpretation. Yet, we also explain the techniques of estimation and inference and how they extend the setting of classical survival analysis with a single event type. We explain some of the more theoretical aspects in separate sections. Two important topics that are not covered in this book are the imputation of missing event types and estimation via the use of pseudo-values.

With respect to the use of R, we assume that one knows how to install a package, how to execute a script, how to select rows and columns, and we assume some familiarity with the use of functions and help files. In the example R code, we write down the full function name because this helps in finding the appropriate help file. For example, we write summary.coxph. Because of R's object orientation, it is sufficient to write summary when performing the analyses.

The book has a website, http://www.competingrisks.org, where you can find additional information. It has a file with hints for making

the computer practicals (ComputerExercisesRHints.pdf) as well as a file with suggested R code, the resulting outcomes and explanatory text (ComputerPracticalsAnswers.pdf). It also contains a script file with all example code that is used in the book (ExampleCode.R) and gives information on upcoming courses.

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I thank the Bikram yoga teachers for their tough classes, which filled me with new energy and creativity. Dear friends, you gave me mental support because you kept asking me about my progress. I hope I have been able to give you some idea of what the book is about. Yet, having finished the book does not mean that I will leave my laptop at home the next time. Jacquelien, I hope to open another bottle with white bubbles soon. Roberto, I have finally finished my "facebookie".



About the Author

Ronald Geskus received his Ph.D. in mathematical statistics at the Delft Technical University in 1997, based on a thesis entitled "Asymptotically efficient estimation with interval censored data" (supervisor Piet Groeneboom). Since 1995 he has been affiliated with the public health service of Amsterdam (PHS), where he performed and supervised studies on HIV and other sexually transmitted infections. There he specialized in the statistical analysis of data collected in cohort studies. In 2014 he was appointed associate professor at the Academic Medical Center (AMC) in Amsterdam. He has worked in several other medical and statistical research environments in the Netherlands and has spent sabbaticals in HIV/AIDS research groups in Paris and Madrid.

His research interests include: i) models for complex time-to-event data (competing risks, multi-state models), ii) models for complex longitudinal data, iii) prediction based on time-updated marker values, iv) causal inference. He contributed to and supervised many medical, epidemiological and statistical studies, performed both at the PHS and the AMC as well as within international collaborations.

He has published around 150 peer reviewed scientific articles, applied as well as methodological. He published methodological papers on i) the estimation of time from HIV infection to AIDS if the time of infection is unknown, ii) the development of markers in relation to disease progression, and iii) the analysis of competing risks with left truncated and right censored data. He is the joint first author of a highly cited tutorial on competing risks and multistate models.

He has been teaching courses on the analysis of competing risks data in Brazil, Spain, Belgium, Sweden, Austria and the Netherlands.



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