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GENERALIZED LINEAR MIXED MODELS

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Preface

When Jim Booth from the University of Florida first approached me about being nominated as the primary lecturer for a Conference Board in the Mathematical Sciences (CBMS) summer conference I was both flattered and slightly apprehensive. When Jim called me to tell me the conference had been approved he only slightly jokingly asked if it was good news or bad. He had to set about organizing a large workshop and I had to prepare ten lectures on the state of generalized linear mixed models (GLMMs). While travelling to the conference and looking at the participant list, I realized there was likely no topic I was lecturing on for which I was the most expert participant. Fortunately, the conference participants were tolerant and engaging and Jim and I both agreed afterwards that it was a wonderful experience and well worthwhile. To a large extent the thanks go to Jim Booth, Jim Hobert, the other speakers (Xihong Lin, David Ruppert, Pat Heagerty and Jim Hobert) and participants.

I apologize for taking so long to get this monograph produced. Shortly after the conference I accepted a position at the University of California, San Francisco. Moving and adjusting to a new position simply did not allow me the time to work on the monograph.

This monograph is a fairly accurate account of the lectures I gave at the CBMS conference with a bit of updating, especially in the bibliographic notes sections that follow each of the later chapters. Compared to my other books, *Variance Components* (with Shayle Searle and George Casella) and *Generalized, Linear, and Mixed Models* (with Shayle Searle) this monograph has much more of a research focus and I have taken the liberty of covering in more depth topics I feel personally passionate about and less obligated to provide textbook-like coverage. Though the title is very similar to *Generalized, Linear, and Mixed Models* the focus is quite different: the lack of commas and the missing “and” are statistically significant!

This monograph is written assuming familiarity with linear models and matrix algebra and some exposure to mixed models and logistic regression. References are given to more standard texts that cover some of the basic material in more depth. The monograph begins with an extended example that introduces all the main ideas. Chapters 2 and 3 briefly review linear mixed and generalized linear models and Chapter 4 defines and introduces GLMMs. Chapter 5 illustrates the breadth of inferential goals possible with GLMMs. One of my main attractions in conducting research on this class of models was the wide variety of practical applications. Chapters 6 through 9 contain the “meat” and tackle the difficult aspects fitting these models to data. The monograph is organized along the lines of

the CBMS lectures.

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