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PREFACE

This is one of two volumes consisting of 33 invited papers presented at the *International Indian Statistical Association Conference* held during October 10–11, 1998, at McMaster University, Hamilton, Ontario, Canada. This Second International Conference of IISA was attended by about 240 participants and included around 170 talks on many different areas of Probability and Statistics. All the papers submitted for publication in this volume were refereed rigorously. The help offered in this regard by the members of the Editorial Board listed earlier and numerous referees is kindly acknowledged. This volume, which includes 33 of the invited papers presented at the conference, focuses on *Advances on Methodological and Applied Aspects of Probability and Statistics*.

For the benefit of the readers, this volume has been divided into nine parts as follows:

Part I	Applied Probability
Part II	Models and Applications
Part III	Estimation and Testing
Part IV	Robust Inference
Part V	Regression and Design
Part VI	Sample Size and Methodology
Part VII	Applications to Industry
Part VIII	Applications to Ecology, Biology and Health
Part IX	Applications to Economics and Management

I sincerely hope that the readers of this volume find the papers to be useful and of interest. I thank all the authors for submitting their papers for publication in this volume.

Special thanks go to Ms. Arnella Moore and Ms. Concetta Seminara-Kennedy of Gordon and Breach Science Publishers for supporting this project and also for helping with the production of this volume. My final thanks go to Mrs. Debbie Iscoe for her fine typesetting of the entire volume.

I hope the readers of this volume enjoy it as much as I did putting it together!

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