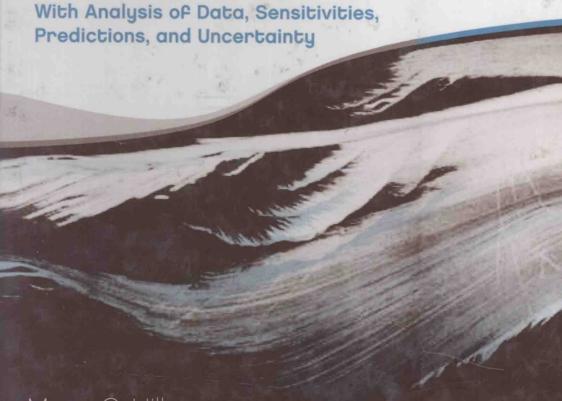


# Effective Groundwater Model Calibration



Mary C. Hill Claire R. Tiedeman

# EFFECTIVE GROUNDWATER MODEL CALIBRATION

With Analysis of Data, Sensitivities, Predictions, and Uncertainty

MARY C. HILL CLAIRE R. TIEDEMAN



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We dedicate this book to the groundwater modelers and software developers of the U.S. Geological Survey. These men and women devote their careers to providing sound scientific analyses for policy makers and to enabling others in the government and the private sector to do the same. We are honored to be their colleagues.

We also dedicate this book to the United States taxpayers, to whom we are ultimately accountable. They have supported our educations, salaries, field work and students. We hope our efforts have improved the understanding and management of their groundwater resources.

With love, we also dedicate this book to our husbands and families.

### **PREFACE**

This book is intended for use in undergraduate and graduate classes, and is also appropriate for use as a reference book and for self-study. Minimal expertise in statistics and mathematics is required for all except a few advanced, optional topics. Knowledge of groundwater principles is needed to understand some parts of the exercises and some of the examples, but students from other fields of science have found classes based on drafts of the book to be very useful.

This book has been more than 12 years in the making. Progressively more mature versions have been used to teach short courses most years since 1991. The short courses have been held at the U.S. Geological Survey National Training Center in Denver, Colorado; the International Ground Water Modeling Center at the Colorado School of Mines in Golden, Colorado; the South Florida Water Management District in West Palm Beach, Florida; the University of Minnesota, in Minneapolis, Minnesota; the Delft University of Technology, The Netherlands; Charles University in Prague, the Czech Republic; University of the Western Cape in Belleville, South Africa; and Utrecht University, The Netherlands. A version also was used to teach a semester course at the University of Colorado in Boulder, Colorado in the fall of 2000. Much of what the book has become results from our many wonderful students. We thank them for their interest, enthusiasm, good humor, and encouragement as we struggled to develop many of the ideas presented in this book.

We also are deeply indebted to the following colleagues for insightful discussions and fruitful collaborations: Richard L. Cooley, Richard M. Yager, Frank A. D'Agnese, Claudia C. Faunt, Arlen W. Harbaugh, Edward R. Banta, Marshall W. Gannett, and D. Matthew Ely of the U.S. Geological Survey, Eileen P. Poeter of the Colorado School of Mines, Evan R. Anderman formerly of Calibra Consultants and McDonald-Morrissey Associates, Inc., Heidi Christiansen Barlebo of the

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Geological Survey of Denmark and Greenland, John Doherty of Watermark Numerical Computing and the University of Queensland (Australia), Karel Kovar of MNP (The Netherlands), Steen Christensen of Aarhus University (Denmark), Theo Olsthoorn of Amsterdam Water Supply (The Netherlands), Richard Waddel of HSI-Geotrans, Inc., Frank Smits formerly of Witteveen + Bos, James Rumbaugh of ESI, Inc., Norm Jones of Utah State University, and Jeff Davis of EMS. In addition, thought-provoking questions from users of MODFLOWP, MODFLOW-2000, PEST, UCODE, and UCODE\_2005 throughout the years have been invaluable.

The book benefited from the careful reviews provided by Peter Kitanidis of Stanford University, Eileen Poeter of the Colorado School of Mines and the International GroundWater Modeling Center (USA), Steen Christensen of the University of Aarhus (Denmark), Roseanna Neupauer of the University of Virginia (USA) (now at the University of Colorado, USA), Luc Lebbe of Ghent University (Belgium), David Lerner of the University of Sheffield (England), Chunmiao Zheng of the University of Alabama (USA), and Howard Reeves and Marshall Gannett of the U.S. Geological Survey. It also benefitted from the kind, professional editors and copyeditor at Wiley: Jonathan Rose, Rosalyn Farkas, and Christina Della Bartolomea.

All errors and omissions are the sole responsibility of the authors.

MARY C. HILL CLAIRE R. TIEDEMAN

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