

Sentiment Analysis

Sentiment analysis is the computational study of people's opinions, sentiments, emotions, moods, and attitudes. This fascinating problem offers numerous research challenges, but promises insights useful to anyone interested in opinion analysis and social media analysis. This comprehensive introduction to the topic takes a natural-language-processing point of view to help readers understand the underlying structure of the problem and the language constructs commonly used to express opinions, sentiments, and emotions. The book covers core areas of sentiment analysis as well as related topics such as debate analysis, intention mining, and fake-opinion detection. It will be a valuable resource for researchers and practitioners in natural language processing, computer science, management sciences, and the social sciences.

In addition to traditional computational methods, this second edition includes recent deep learning methods to analyze and summarize sentiments and opinions, and also new material on emotion and mood analysis techniques, emotion-enhanced dialogues, and multimodal emotion analysis.

BING LIU is a distinguished professor of computer science at the University of Illinois at Chicago. His current research interests include sentiment analysis, lifelong machine learning, natural language processing, and data mining. He has published extensively in top conferences and journals, and his research has been cited on the front page of *The New York Times*. Three of his research papers also received Test-of-Time awards. He was the recipient of the ACM SIGKDD Innovation Award in 2018, and he is a Fellow of the ACM, AAAI, and IEEE. He served as the Chair of ACM SIGKDD from 2013 to 2017.

Studies in Natural Language Processing

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Sentiment Analysis

Mining Opinions, Sentiments, and Emotions

Second Edition

Bing Liu

University of Illinois at Chicago, and Peking University



CAMBRIDGE
UNIVERSITY PRESS

CAMBRIDGE UNIVERSITY PRESS

University Printing House, Cambridge CB2 8BS, United Kingdom

One Liberty Plaza, 20th Floor, New York, NY 10006, USA

477 Williamstown Road, Port Melbourne, VIC 3207, Australia

314–321, 3rd Floor, Plot 3, Splendor Forum, Jasola District Centre,
New Delhi – 110025, India

79 Anson Road, #06–04/06, Singapore 079906

Cambridge University Press is part of the University of Cambridge.

It furthers the University's mission by disseminating knowledge in the pursuit of education, learning, and research at the highest international levels of excellence.

www.cambridge.org

Information on this title: www.cambridge.org/9781108486378

DOI: 10.1017/9781108639286

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Second edition 2020

First published 2015

Printed in the United Kingdom by TJ International, Padstow Cornwall

A catalogue record for this publication is available from the British Library.

Library of Congress Cataloging-in-Publication Data

Names: Liu, Bing, 1963– author.

Title: Sentiment analysis : mining opinions, sentiments, and emotions /
Bing Liu, University of Illinois, Chicago.

Description: Second edition. | Cambridge ; New York : Cambridge University Press, 2020. | Series: Studies in natural language processing | Includes bibliographical references and index.

Identifiers: LCCN 2020009497 (print) | LCCN 2020009498 (ebook) |

ISBN 9781108486378 (hardback) | ISBN 9781108639286 (epub)

Subjects: LCSH: Natural language processing (Computer science) |

Computational linguistics. | Public opinion–Data processing. | Data mining. |

Discourse analysis–Data processing. | Language and emotions.

Classification: LCC QA76.9.N38 L58 2020 (print) | LCC QA76.9.N38 (ebook) |

DDC 006.3/12–dc23

LC record available at <https://lcn.loc.gov/2020009497>

LC ebook record available at <https://lcn.loc.gov/2020009498>

ISBN 978-1-108-48637-8 Hardback

Cambridge University Press has no responsibility for the persistence or accuracy of URLs for external or third-party internet websites referred to in this publication and does not guarantee that any content on such websites is, or will remain, accurate or appropriate.

Contents

<i>Preface</i>	<i>page xi</i>
<i>Acknowledgments</i>	<i>xvi</i>
1 Introduction	1
1.1 Sentiment Analysis Applications	5
1.2 Sentiment Analysis Research	9
1.2.1 Different Levels of Analysis	9
1.2.2 Sentiment Lexicon and Its Issues	11
1.2.3 Analyzing Debates and Comments	12
1.2.4 Mining Intent	13
1.2.5 Opinion Spam Detection and Quality of Reviews	14
1.3 Sentiment Analysis As Mini-NLP	15
1.4 My Approach to Writing This Book	16
2 The Problem of Sentiment Analysis	18
2.1 Definition of Opinion	19
2.1.1 Opinion Definition	20
2.1.2 Sentiment Target	21
2.1.3 Sentiment of Opinion	22
2.1.4 Opinion Definition Simplified	24
2.1.5 Reason and Qualifier for Opinion	26
2.1.6 Objective and Tasks of Sentiment Analysis	28
2.2 Definition of Opinion Summary	32
2.3 Affect, Emotion, and Mood	34
2.3.1 Affect, Emotion, and Mood in Psychology	35
2.3.2 Emotion	37
2.3.3 Mood	40
2.3.4 Feeling	41
2.3.5 Affect, Emotion, and Mood in Sentiment Analysis	43
2.4 Different Types of Opinions	46
2.4.1 Regular and Comparative Opinions	46
2.4.2 Subjective and Fact-Implied Opinions	47
2.4.3 First-Person and Non-First-Person Opinions	51
2.4.4 Meta-Opinions	52
2.5 Author and Reader Standpoint	52
2.6 Summary	53

vi	Contents	
3	Document Sentiment Classification	55
3.1	Supervised Sentiment Classification	57
3.1.1	Classification Using Traditional Machine Learning Algorithms	57
3.1.2	Classification Using a Custom Score Function	66
3.1.3	Classification Using Deep Learning	67
3.1.4	Classification Based on Lifelong Learning	70
3.2	Unsupervised Sentiment Classification	74
3.2.1	Classification Using Syntactic Patterns and Web Search	74
3.2.2	Classification Using Sentiment Lexicons	76
3.3	Sentiment Rating Prediction	79
3.4	Cross-Domain Sentiment Classification	81
3.5	Cross-Language Sentiment Classification	84
3.6	Emotion Classification of Documents	86
3.7	Summary	88
4	Sentence Subjectivity and Sentiment Classification	89
4.1	Subjectivity	91
4.2	Sentence Subjectivity Classification	92
4.3	Sentence Sentiment Classification	96
4.3.1	Assumption of Sentence Sentiment Classification	96
4.3.2	Traditional Classification Methods	97
4.3.3	Deep Learning–Based Methods	99
4.4	Dealing with Conditional Sentences	102
4.5	Dealing with Sarcastic Sentences	104
4.6	Cross-Language Subjectivity and Sentiment Classification	107
4.7	Using Discourse Information for Sentiment Classification	109
4.8	Emotion Classification of Sentences	110
4.9	Multimodal Sentiment and Emotion Classification	112
4.10	Summary	113
5	Aspect Sentiment Classification	115
5.1	Aspect Sentiment Classification	116
5.1.1	Supervised Learning	117
5.1.2	Lexicon-Based Approach	121
5.1.3	Pros and Cons of the Two Approaches	125
5.2	Rules of Sentiment Composition	126
5.2.1	Sentiment Composition Rules	128
5.2.2	DECREASE and INCREASE Expressions	135
5.2.3	SMALL_OR_LESS and LARGE_OR_MORE Expressions	138
5.2.4	Emotion and Sentiment Intensity	141
5.2.5	Senses of Sentiment Words	142
5.2.6	Survey of Other Approaches	144
5.3	Negation and Sentiment	146
5.3.1	Negation Words	146
5.3.2	Never	149
5.3.3	Some Other Common Sentiment Shifters	151
5.3.4	Shifted or Transferred Negations	152
5.3.5	Scope of Negations	152

Contents	vii
5.4 Modality and Sentiment	153
5.5 Coordinating Conjunction <i>But</i>	158
5.6 Sentiment Words in Non-Opinion Contexts	160
5.7 Rule Representation	162
5.8 Word Sense Disambiguation and Coreference Resolution	164
5.9 Summary	166
6 Aspect and Entity Extraction	168
6.1 Frequency-Based Aspect Extraction	169
6.2 Exploiting Syntactic Relations	171
6.2.1 Using Opinion and Target Relations	172
6.2.2 Using Part-of and Attribute-of Relations	179
6.3 Using Supervised Learning	182
6.3.1 Hidden Markov Model	182
6.3.2 Conditional Random Fields	183
6.3.3 Deep Learning-Based Methods	186
6.4 Mapping Implicit Aspects	188
6.4.1 Corpus-Based Approach	188
6.4.2 Dictionary-Based Approach	189
6.5 Grouping Aspects into Categories	192
6.6 Exploiting Topic Models	194
6.6.1 Latent Dirichlet Allocation	196
6.6.2 Using Unsupervised Topic Models	199
6.6.3 Using Prior Domain Knowledge in Modeling	205
6.6.4 Lifelong Topic Models: Learn As Humans Do	207
6.6.5 Using Phrases As Topical Terms	211
6.7 Entity Extraction and Resolution	216
6.7.1 The Problem of Entity Extraction and Resolution	217
6.7.2 Entity Extraction	220
6.7.3 Entity Linking	222
6.7.4 Entity Search and Linking	224
6.8 Opinion Holder and Time Extraction	224
6.9 Summary	225
7 Sentiment Lexicon Generation	227
7.1 Dictionary-Based Approach	228
7.2 Corpus-Based Approach	232
7.2.1 Identifying Sentiment Words from a Corpus	232
7.2.2 Dealing with Context-Dependent Sentiment Words	233
7.2.3 Lexicon Adaptation	236
7.2.4 Some Other Related Work	237
7.3 Sentiment Word Embedding	238
7.4 Desirable and Undesirable Facts	239
7.5 Summary	241
8 Analysis of Comparative Opinions	243
8.1 Problem Definition	243
8.2 Identifying Comparative Sentences	247
8.3 Identifying the Preferred Entity Set	248
8.4 Special Types of Comparisons	250

viii	Contents	
	8.4.1 Nonstandard Comparisons	250
	8.4.2 Cross-Type Comparison	253
	8.4.3 Single-Entity Comparison	254
	8.4.4 Sentences Involving <i>Compare</i> or <i>Comparison</i>	255
	8.5 Entity and Aspect Extraction	257
	8.6 Summary	258
9	Opinion Summarization and Search	259
	9.1 Aspect-Based Opinion Summarization	260
	9.2 Enhancements to Aspect-Based Summaries	262
	9.3 Contrastive View Summarization	266
	9.4 Traditional Summarization	266
	9.5 Summarization of Comparative Opinions	267
	9.6 Opinion Search	267
	9.7 Existing Opinion Retrieval Techniques	269
	9.8 Summary	271
10	Analysis of Debates and Comments	273
	10.1 Recognizing Stances in Debates	274
	10.2 Modeling Debates/Discussions	277
	10.2.1 JTE Model	279
	10.2.2 JTE-R Model: Encoding Reply Relations	284
	10.2.3 JTE-P Model: Encoding Pair Structures	286
	10.2.4 Analysis of Tolerance in Online Discussions	288
	10.3 Modeling Comments	290
	10.4 Summary	292
11	Mining Intent	294
	11.1 The Problem of Intent Mining	294
	11.2 Intent Classification	298
	11.3 Fine-Grained Mining of Intent	301
	11.4 Summary	302
12	Detecting Fake or Deceptive Opinions	304
	12.1 Different Types of Spam	307
	12.1.1 Harmful Fake Reviews	308
	12.1.2 Types of Spammers and Spamming	309
	12.1.3 Types of Data, Features, and Detection	311
	12.1.4 Fake Reviews versus Conventional Lies	313
	12.2 Supervised Fake Review Detection	315
	12.3 Supervised Yelp Data Experiment	318
	12.3.1 Supervised Learning Using Linguistic Features	319
	12.3.2 Supervised Learning Using Behavioral Features	321
	12.4 Automated Discovery of Abnormal Patterns	322
	12.4.1 Class Association Rules	322
	12.4.2 Unexpectedness of One-Condition Rules	324
	12.4.3 Unexpectedness of Two-Condition Rules	327
	12.5 Model-Based Behavioral Analysis	329
	12.5.1 Spam Detection Based on Atypical Behaviors	330
	12.5.2 Spam Detection Using Review Graphs	331

Contents	ix
12.5.3 Spam Detection Using Bayesian Models	332
12.6 Group Spam Detection	333
12.6.1 Group Behavior Features	337
12.6.2 Individual Member Behavior Features	340
12.7 Identifying Reviewers with Multiple Userids	341
12.7.1 Learning in a Similarity Space	342
12.7.2 Training Data Preparation	343
12.7.3 <i>d</i> -Features and <i>s</i> -Features	344
12.7.4 Identifying Userids of the Same Author	345
12.8 Exploiting Burstiness in Reviews	348
12.9 Future Research Directions	351
12.10 Summary	352
13 Quality of Reviews	354
13.1 Quality Prediction As a Regression Problem	354
13.2 Other Methods	356
13.3 Some New Frontiers	358
13.4 Summary	359
14 Conclusion	360
<i>Appendix</i>	365
<i>Bibliography</i>	376
<i>Index</i>	427

Cambridge University Press
978-1-108-48637-8 — Sentiment Analysis
2nd Edition
Frontmatter
[More Information](#)

Preface

Since the first edition of the book was published in 2015, we have witnessed the rapid rise of deep learning, which has resulted in a huge amount of work on using deep learning models to perform sentiment analysis tasks. The goal of this second edition is mainly to update the first edition with the deep learning methods published in the past few years. In addition, many other sections of the book have been updated.

Opinion and sentiment and their related concepts, such as evaluation, appraisal, attitude, affect, emotion, and mood, are about our subjective feelings and beliefs. They are central to human psychology and are key influencers of our behaviors. Our beliefs and perceptions of reality, as well as the choices we make, are to a considerable degree conditioned on how others see and perceive the world. For this reason, our views of the world are very much influenced by others' views. Whenever we need to make a decision, we often seek out others' opinions. This is true not only for individuals but also for organizations. From an application point of view, we naturally want to mine people's opinions and feelings toward any subject matter of interest, which is the task of sentiment analysis. More precisely, sentiment analysis – also called opinion mining – is a field of study that aims to extract opinions and sentiments from natural language text using computational methods.

The inception and rapid growth of sentiment analysis coincide with those of social media on the web, such as reviews, forum discussions, blogs, and microblogs, because for the first time in human history, we have a huge volume of opinionated data recorded in digital forms. These data, also called *user-generated content*, prompted researchers to mine them to discover useful knowledge. This endeavor naturally led to the problem of sentiment analysis or opinion mining because these data are full of opinions. This is not surprising, because the primary reason that people post messages on social media platforms is to express their views and opinions; therefore, sentiment analysis is at the very core of social media analysis. Since the early 2000s, sentiment analysis has grown to be one of the most active research areas in natural language processing. It is also widely studied in the context of data mining,

web mining, and information retrieval. In fact, this line of research has expanded from computer science to management science and social science because of its importance to business and society as a whole. Industrial activities surrounding sentiment analysis have also thrived. Numerous start-ups have emerged. Many large corporations – for example, Microsoft, Google, Facebook, Hewlett-Packard, IBM, Adobe, Alibaba, Baidu, and Tencent – have also built their own in-house systems. Sentiment analysis systems have found applications in almost every business, health, government, and social domain.

Although no silver bullet algorithm can solve the sentiment analysis problem, many deployed systems are able to provide useful information to support real-life applications. I believe it is now a good time to document the knowledge that we have obtained through research – and, to some extent, in practice – in a book. Obviously, I don't claim to know everything that is happening in the industry, as businesses do not publish or disclose their algorithms. However, I have built a sentiment analysis system myself in a start-up company and served clients on projects involving social media data sets in a large variety of domains. Over the years, many developers of sentiment analysis systems in the industry have also told me roughly what algorithms they were using. Thus, I have a reasonable knowledge of practical systems and their capabilities and firsthand experience in solving real-life problems. I try to pass along those nonconfidential pieces of information and knowledge in this book.

In writing this book, I aimed to take a balanced approach, analyzing the sentiment analysis problem from a linguistic angle to help readers understand the underlying structure of the problem and the language constructs commonly used to express opinions and sentiments, and presenting computational methods to analyze and summarize opinions. Like many natural language processing tasks, most published computational techniques use machine learning or data mining algorithms with the help of text-specific clues or features. However, if we focus solely on such computational algorithms, we will miss the deep insights of the problem, which in turn will hinder our progress on the computational front. Most existing machine learning algorithms are black boxes; they do not produce human-interpretable models. In turn, when something goes wrong, it is hard to know the cause and how to fix it.

In presenting linguistic constructs and perspectives, I do not follow the linguistic tradition in writing, because the knowledge and the way that the knowledge is presented in the traditional linguistics literature are mainly for people to understand, rather than for computers to operationalize to solve real-life problems. Although the knowledge of human beings and instructions for computers can largely intersect, they also have major differences. As a case in point, when I was working on the problem of mining opinions from conditional sentences, I read several linguistics books about conditionals. To my

surprise, I found almost no linguistic knowledge that could be operationalized computationally to help solve the problem. I believe this is partly because the current computation technologies are not mature enough to have the same understanding capability as people and partly because much of the linguistic knowledge was not meant for computers to use.

Another feature of this book is that it is not just about studying the language for human understanding per se, as much of the traditional linguistic literature is; it is also about practical applications of mining sentiment and opinion expressed in natural language, for which we want not only to recognize sentiment or opinion expressions and their polarities (or orientations) but also to extract several other pieces of important information associated with sentiment or opinion. For example, we want to identify the real-world entities or topics that a sentiment or opinion is about. These entities or topics are called *opinion* (or *sentiment*) *targets*. Extracting opinion targets is extremely important in practice. For example, in the sentence “I am disgusted by the tax increase for the poor,” if we find only that the sentence expresses a negative sentiment and/or the emotion of *disgust* from the sentence author, it is not that useful in practice. But if we also find that the negative sentiment is toward “tax increase for the poor,” which is the target of the negative sentiment or emotion, the information becomes much more valuable. I hope this book can also serve to encourage linguists to develop a comprehensive theory about sentiment, opinion, and emotion, as well as their associated concepts.

I wrote this book as an introductory text for the field of sentiment analysis and as a research survey. In many places, it is one or the other; in other places, it is a mixture of both. The reason for this mixed or somewhat unusual presentational style is that there are few truly mature techniques or algorithms for sentiment analysis, although numerous researchers have attempted to solve each subproblem using many techniques. In many cases, we can see from the accuracy of results of the published papers that they are not yet ready for prime time.

Another reason for the mixed presentational style of this book is that most existing research methods are direct applications of machine learning and data mining algorithms employing text features or the raw text itself. Because many books on machine learning and data mining cover these algorithms extensively, the algorithms are not detailed in this book. This book also does not detail the basics of linguistics or natural language processing, such as part-of-speech tagging, syntactic parsing, shallow parsing, and grammar. Although these topics are very important to sentiment analysis, once again they have been covered in numerous books on natural language processing. This book thus assumes that readers know the basics of machine learning and natural language processing.

I tried to cover all major developments of the field in this book. It is thus quite comprehensive. Evidence of this is that the book cites more than seven

hundred publications from all major conferences and journals. I organized the book as follows.

Chapter 1 introduces the book and gives the motivations for the study of sentiment analysis. We see that sentiment analysis is a fascinating, yet challenging problem with almost unlimited practical applications.

Chapter 2 defines the sentiment analysis problem and discusses many of its related issues. Here we see that although sentiment analysis is a natural language processing problem, it can be defined structurally. Through this definition, we can transform unstructured text to structured data. This facilitates subsequent qualitative and quantitative analyses, which are critical for real-life applications. We also see that sentiment analysis is a multifaceted problem with many challenging and interrelated subproblems.

Chapter 3 studies the popular topic of document-level sentiment classification, which classifies an opinion document (e.g., a product review) as expressing a positive or negative opinion. Chapter 4 studies the same classification problem but focuses on each individual sentence. Related problems of sentiment rating prediction, transfer learning, and multilingual sentiment classification are also discussed in these two chapters.

Chapters 5 and 6 go to a fine-grained level to study the enormously important topic of aspect-based sentiment analysis, which not only classifies sentiment but also identifies the target of sentiment or opinion. Most practical sentiment analysis or opinion mining systems in industry are based on this fine-grained level of analysis. Chapter 5 focuses on aspect sentiment classification, and Chapter 6 addresses aspect or target extraction.

Chapter 7 describes research that compiles sentiment lexicons. A sentiment lexicon is a list of words and phrases (e.g., *good*, *amazing*, *bad*, *horrible*) that people often use to express positive or negative opinions. Chapter 8 studies opinions expressed in comparative sentences. Chapter 9 focuses on opinion summarization and opinion search. Chapter 10 looks into a different type of sentiment (agreement and disagreement) expressed in online debates and discussions, which involve extensive interactive exchanges among participants. Chapter 11 investigates intent mining, which aims to discover intentions expressed in language.

Chapter 12 switches to a very different topic: the detection of fake or deceptive online opinions. Chapter 13 studies the problem of ranking online reviews based on their usefulness so that users can view the most useful reviews first. Chapter 14 concludes the book and discusses some future research.

This book is suitable for students, researchers, and practitioners who are interested in social media analysis and natural language processing in general, and sentiment analysis or opinion mining in particular. It is written not only for the computer science audience, but also for researchers and practitioners in

management sciences and social sciences. Consumer sentiments and public opinions are central to many management and social science areas such as marketing, economics, communication, political science, and even history. Lecturers can readily use the book in class for courses on natural language processing, social media analysis, social computing, and text and data mining. Lecture slides are available online.

Acknowledgments

Many researchers assisted me technically while I was writing this book. Without their help, the book might never have become reality. First of all, I would like to thank my former and current students – Zhiyuan Chen, Junsheng Cheng, Xiaowen Ding, Geli Fei, Murthy Ganapathibhotla, Mingqing Hu, Wenpeng Hu (Peking University), Nitin Jindal, Zixuan Ke, Gyuhak Kim, Abhinav Kumar, Huayi Li, Guangyi Lv (visiting student from University of Science and Technology China), Nianzu Ma, Sahisnu Mazumder, Arjun Mukherjee, Ramanathan Narayanan (Northwestern University), Qi Qin (Peking University), Guang Qiu (visiting student from Zhejiang University), Sathish Ramadoss, Lei Shu, Jianfeng Si (visiting student from City University of Hong Kong), Vivek Venkataraman, William Underwood, Andrea Vaccari, Hao Wang (visiting student from Southwest Jiaotong University), Shuai Wang, Hu Xu, Zhongwu Zhai (visiting student from Tsinghua University), and Lei Zhang – for contributing numerous research ideas over the years. Discussions with many researchers also helped shape the book: Shuanhu Bai, Jim Blomo, Malu G. Castellanos, Dennis Chong, Umesh Dayal, Eduard Dragut, Boi Faltings, Ronen Feldman, Christiane D. Fellbaum, Zhiqiang Gao, Riddhiman Ghosh, Natalie Glance, Meichun Hsu, Joshua Huang, Minglie Huang, Jing Jiang, Birgit König, Xiao-li Li, Qian Liu, Boia Marina, Sharon Meraz, Tieyun Qian, Jidong Shao, Mehrbod Sharifi, Hao Wang, Jan Wiebe, Qiang Yang, Lixia Yao, Clement Yu, Philip S. Yu, ChengXiang Zhai, Fangwei Zhang, Yuanlin Zhang, Jun Zhao, Xiaoyan Zhu, and Chengqing Zong. Xueying Zhang from Peking University helped greatly tidy up and correct a large number of references. I am also in debt to the two anonymous reviewers; despite their busy schedules, they read the book very carefully and gave me numerous excellent suggestions. I have taken each and every one of those recommendations into consideration while improving this book. Both the web and social media made writing this book so much easier: I have found a great deal of valuable information from them that tremendously helped me in writing the book.

I would also like to express my gratitude to the National Science Foundation, Google, HP Labs, Tencent Holdings, Huawei Technologies, Bosch, and

Microsoft Corporation for their generous support of my research over the years. The Department of Computer Science, University of Illinois at Chicago (UIC), and Wangxuan Institute of Computing Technology, Peking University, provided computing resources and very supportive environments for this project. The new materials of this second edition was mainly written when I was on leave from UIC at Peking University.

On the publication side, it was a pleasure working with the helpful staff of Cambridge University Press. I thank my editors, Lauren Cowles (first edition), Amy He, and Kaitlin Leach (second edition). It has been a wonderful experience working with them. I also thank my copy editor, Holly T. Monteith (first edition) and Jill E. Hobbs (second edition), for helping me improve the presentation and my production editors, Sonika Rai (first edition) and Mathew Rohit (second edition), for guiding me through the final production process.

Finally, I thank my parents, brother, and sister for their constant support and encouragement. My deepest gratitude goes to my own family: Yue, Shelley, and Kate. They have helped me in so many ways. My wife has taken care of almost everything at home and put up with me and the long hours that I have spent on this book. I dedicate this book to them.

Cambridge University Press
978-1-108-48637-8 — Sentiment Analysis
2nd Edition
Frontmatter
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