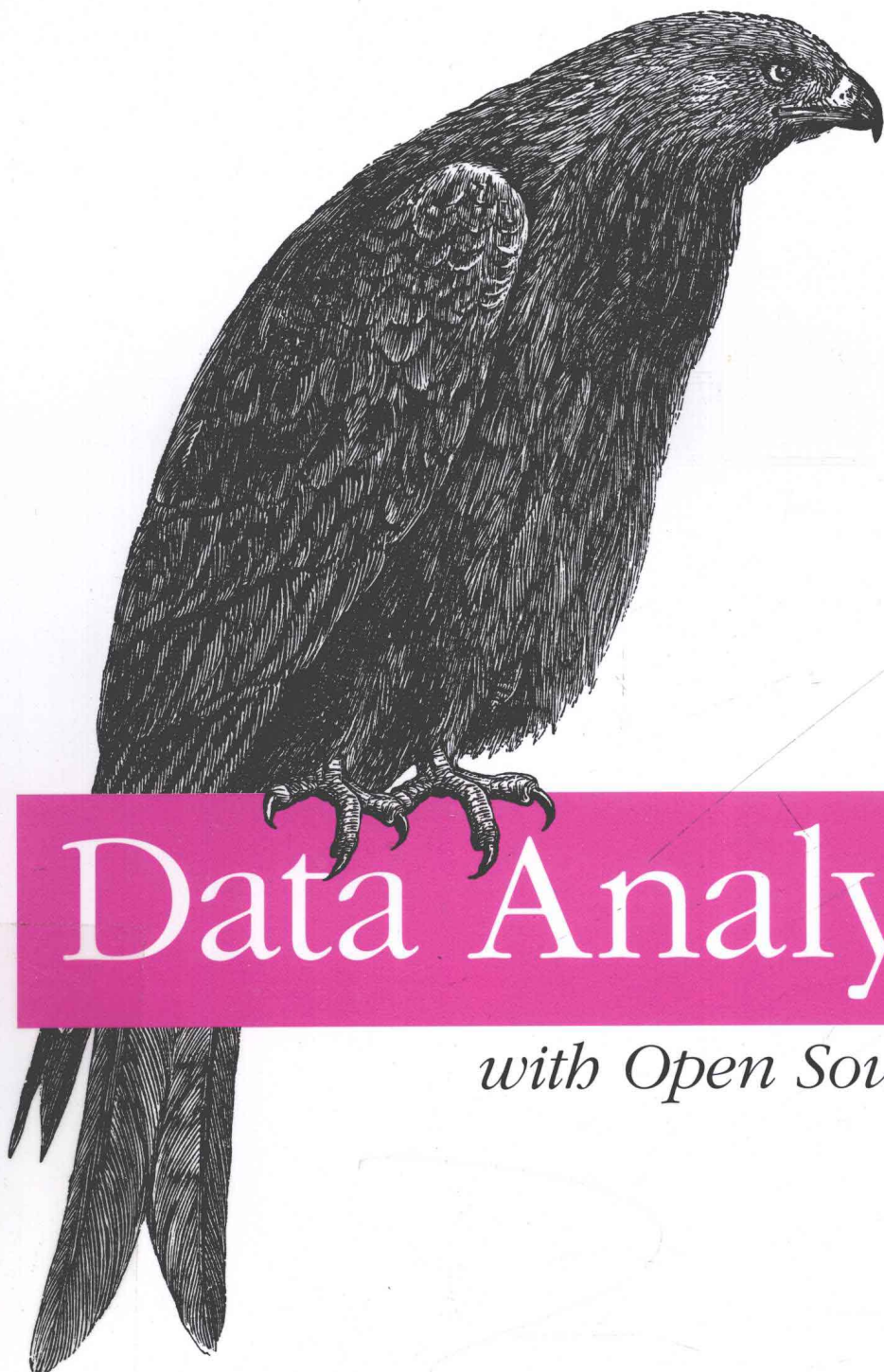


基于开源工具的数据分析 (影印版)



Data Analysis

with Open Source Tools

O'REILLY®

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Philipp K. Janert 著

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Data Analysis with Open Source Tools

Philipp K. Janert

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基于开源工具的数据分析(影印版)

Data Analysis with Open Source Tools

Furious activity is no substitute for understanding.

—H. H. Williams

Preface

THIS BOOK GREW OUT OF MY EXPERIENCE OF WORKING WITH DATA FOR VARIOUS COMPANIES IN THE TECH industry. It is a collection of those concepts and techniques that I have found to be the most useful, including many topics that I wish I had known earlier—but didn't.

My degree is in physics, but I also worked as a software engineer for several years. The book reflects this dual heritage. On the one hand, it is written for programmers and others in the software field: I assume that you, like me, have the ability to write your own programs to manipulate data in any way you want.

On the other hand, the way I think about data has been shaped by my background and education. As a physicist, I am not content merely to describe data or to make black-box predictions: the purpose of an analysis is always to develop an understanding for the processes or mechanisms that give rise to the data that we observe.

The instrument to express such understanding is the *model*: a description of the system under study (in other words, not just a description of the data!), simplified as necessary but nevertheless capturing the relevant information. A model may be crude ("Assume a spherical cow . . ."), but if it helps us develop better insight on how the system works, it is a successful model nevertheless. (Additional precision can often be obtained at a later time, if it is really necessary.)

This emphasis on models and simplified descriptions is not universal: other authors and practitioners will make different choices. But it is essential to my approach and point of view.

This is a rather personal book. Although I have tried to be reasonably comprehensive, I have selected the topics that I consider relevant and useful in practice—whether they are part of the "canon" or not. Also included are several topics that you won't find in any other book on data analysis. Although neither new nor original, they are usually not used or discussed in this particular context—but I find them indispensable.

Throughout the book, I freely offer specific, explicit advice, opinions, and assessments. These remarks are reflections of my personal interest, experience, and understanding. I do not claim that my point of view is necessarily correct: evaluate what I say for yourself and feel free to adapt it to your needs. In my view, a specific, well-argued position is of greater use than a sterile laundry list of possible algorithms—even if you later decide to disagree with me. The value is not in the opinion but rather in the arguments leading up to it. If your arguments are better than mine, or even just more agreeable to you, then I will have achieved my purpose!

Data analysis, as I understand it, is not a fixed set of techniques. It is a way of life, and it has a name: curiosity. There is always something else to find out and something more to learn. This book is not the last word on the matter; it is merely a snapshot in time: things I knew about and found useful today.

“Works are of value only if they give rise to better ones.”

(Alexander von Humboldt, writing to Charles Darwin, 18 September 1839)

Before We Begin

More data analysis efforts seem to go bad because of an excess of sophistication rather than a lack of it.

This may come as a surprise, but it has been my experience again and again. As a consultant, I am often called in when the initial project team has already gotten stuck. Rarely (if ever) does the problem turn out to be that the team did not have the required skills. On the contrary, I usually find that they tried to do something unnecessarily complicated and are now struggling with the consequences of their own invention!

Based on what I have seen, two particular risk areas stand out:

- The use of “statistical” concepts that are only partially understood (and given the relative obscurity of most of statistics, this includes virtually *all* statistical concepts)
- Complicated (and expensive) black-box solutions when a simple and transparent approach would have worked at least as well or better

I strongly recommend that you make it a habit to avoid all statistical language. Keep it simple and stick to what you know for sure. There is absolutely nothing wrong with speaking of the “range over which points spread,” because this phrase means exactly what it says: the range over which points spread, and only that! Once we start talking about “standard deviations,” this clarity is gone. Are we still talking about the *observed* width of the distribution? Or are we talking about one specific *measure* for this width? (The standard deviation is only one of several that are available.) Are we already making an implicit *assumption* about the nature of the distribution? (The standard deviation is only suitable under certain conditions, which are often not fulfilled in practice.) Or are we even confusing the *predictions* we could make if these assumptions were true with the actual data? (The moment someone talks about “95 percent anything” we know it’s the latter!)

I’d also like to remind you not to discard simple methods until they have been *proven* insufficient. Simple solutions are frequently rather effective: the marginal benefit that more complicated methods can deliver is often quite small (and may be in no reasonable relation to the increased cost). More importantly, simple methods have fewer opportunities to go wrong or to obscure the obvious.

True story: a company was tracking the occurrence of defects over time. Of course, the actual number of defects varied quite a bit from one day to the next, and they were looking for a way to obtain an estimate for the typical number of expected defects. The solution proposed by their IT department involved a compute cluster running a neural network! (I am not making this up.) In fact, a one-line calculation (involving a moving average or single exponential smoothing) is all that was needed.

I think the primary reason for this tendency to make data analysis projects more complicated than they are is *discomfort*: discomfort with an unfamiliar problem space and uncertainty about how to proceed. This discomfort and uncertainty creates a desire to bring in the “big guns”: fancy terminology, heavy machinery, large projects. In reality, of course, the opposite is true: the complexities of the “solution” overwhelm the original problem, and nothing gets accomplished.

Data analysis does not have to be all that hard. Although there are situations when elementary methods will no longer be sufficient, they are much less prevalent than you might expect. In the vast majority of cases, curiosity and a healthy dose of common sense will serve you well.

The attitude that I am trying to convey can be summarized in a few points:

- Simple is better than complex.
- Cheap is better than expensive.
- Explicit is better than opaque.
- Purpose is more important than process.
- Insight is more important than precision.
- Understanding is more important than technique.
- Think more, work less.

Although I do acknowledge that the items on the right are necessary at times, I will give preference to those on the left whenever possible.

It is in this spirit that I am offering the concepts and techniques that make up the rest of this book.

Conventions Used in This Book

The following typographical conventions are used in this book:

Italic

Indicates new terms, URLs, and email addresses

`Constant width`

Used to refer to language and script elements

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During the preparation of this book, the excellent collection at the University of Washington libraries was an especially valuable resource to me.

Authors usually thank their spouses for their “patience and support” or words to that effect. Unless one has lived through the actual experience, one cannot fully comprehend how true this is. Over the last three years, Angela has endured what must have seemed like a nearly continuous stream of whining, frustration, and desperation—punctuated by occasional outbursts of exhilaration and grandiosity—all of which before the background of the self-centered and self-absorbed attitude of a typical author. Her patience and support were unfailing. It’s her turn now.

CONTENTS

PREFACE	xiii
1 INTRODUCTION	1
<i>Data Analysis</i>	1
<i>What's in This Book</i>	2
<i>What's with the Workshops?</i>	3
<i>What's with the Math?</i>	4
<i>What You'll Need</i>	5
<i>What's Missing</i>	6
 PART I Graphics: Looking at Data	
2 A SINGLE VARIABLE: SHAPE AND DISTRIBUTION	11
<i>Dot and Jitter Plots</i>	12
<i>Histograms and Kernel Density Estimates</i>	14
<i>The Cumulative Distribution Function</i>	23
<i>Rank-Order Plots and Lift Charts</i>	30
<i>Only When Appropriate: Summary Statistics and Box Plots</i>	33
<i>Workshop: NumPy</i>	38
<i>Further Reading</i>	45
3 TWO VARIABLES: ESTABLISHING RELATIONSHIPS	47
<i>Scatter Plots</i>	47
<i>Conquering Noise: Smoothing</i>	48
<i>Logarithmic Plots</i>	57
<i>Banking</i>	61
<i>Linear Regression and All That</i>	62
<i>Showing What's Important</i>	66
<i>Graphical Analysis and Presentation Graphics</i>	68
<i>Workshop: matplotlib</i>	69
<i>Further Reading</i>	78
4 TIME AS A VARIABLE: TIME-SERIES ANALYSIS	79
<i>Examples</i>	79
<i>The Task</i>	83
<i>Smoothing</i>	84
<i>Don't Overlook the Obvious!</i>	90
<i>The Correlation Function</i>	91

	<i>Optional: Filters and Convolutions</i>	95
	<i>Workshop: <code>scipy.signal</code></i>	96
	<i>Further Reading</i>	98
5	MORE THAN TWO VARIABLES: GRAPHICAL MULTIVARIATE ANALYSIS	99
	<i>False-Color Plots</i>	100
	<i>A Lot at a Glance: Multiplots</i>	105
	<i>Composition Problems</i>	110
	<i>Novel Plot Types</i>	116
	<i>Interactive Explorations</i>	120
	<i>Workshop: Tools for Multivariate Graphics</i>	123
	<i>Further Reading</i>	125
6	INTERMEZZO: A DATA ANALYSIS SESSION	127
	<i>A Data Analysis Session</i>	127
	<i>Workshop: <code>gnuplot</code></i>	136
	<i>Further Reading</i>	138

PART II Analytics: Modeling Data

7	GUESSTIMATION AND THE BACK OF THE ENVELOPE	141
	<i>Principles of Guesstimation</i>	142
	<i>How Good Are Those Numbers?</i>	151
	<i>Optional: A Closer Look at Perturbation Theory and Error Propagation</i>	155
	<i>Workshop: The Gnu Scientific Library (GSL)</i>	158
	<i>Further Reading</i>	161
8	MODELS FROM SCALING ARGUMENTS	163
	<i>Models</i>	163
	<i>Arguments from Scale</i>	165
	<i>Mean-Field Approximations</i>	175
	<i>Common Time-Evolution Scenarios</i>	178
	<i>Case Study: How Many Servers Are Best?</i>	182
	<i>Why Modeling?</i>	184
	<i>Workshop: Sage</i>	184
	<i>Further Reading</i>	188
9	ARGUMENTS FROM PROBABILITY MODELS	191
	<i>The Binomial Distribution and Bernoulli Trials</i>	191
	<i>The Gaussian Distribution and the Central Limit Theorem</i>	195
	<i>Power-Law Distributions and Non-Normal Statistics</i>	201
	<i>Other Distributions</i>	206
	<i>Optional: Case Study—Unique Visitors over Time</i>	211
	<i>Workshop: Power-Law Distributions</i>	215
	<i>Further Reading</i>	218

10	WHAT YOU REALLY NEED TO KNOW ABOUT CLASSICAL STATISTICS	221
	<i>Genesis</i>	221
	<i>Statistics Defined</i>	223
	<i>Statistics Explained</i>	226
	<i>Controlled Experiments Versus Observational Studies</i>	230
	<i>Optional: Bayesian Statistics—The Other Point of View</i>	235
	<i>Workshop: R</i>	243
	<i>Further Reading</i>	249
11	INTERMEZZO: MYTHBUSTING—BIGFOOT, LEAST SQUARES, AND ALL THAT	253
	<i>How to Average Averages</i>	253
	<i>The Standard Deviation</i>	256
	<i>Least Squares</i>	260
	<i>Further Reading</i>	264
 PART III Computation: Mining Data		
12	SIMULATIONS	267
	<i>A Warm-Up Question</i>	267
	<i>Monte Carlo Simulations</i>	270
	<i>Resampling Methods</i>	276
	<i>Workshop: Discrete Event Simulations with SimPy</i>	280
	<i>Further Reading</i>	291
13	FINDING CLUSTERS	293
	<i>What Constitutes a Cluster?</i>	293
	<i>Distance and Similarity Measures</i>	298
	<i>Clustering Methods</i>	304
	<i>Pre- and Postprocessing</i>	311
	<i>Other Thoughts</i>	314
	<i>A Special Case: Market Basket Analysis</i>	316
	<i>A Word of Warning</i>	319
	<i>Workshop: Pycluster and the C Clustering Library</i>	320
	<i>Further Reading</i>	324
14	SEEING THE FOREST FOR THE TREES: FINDING IMPORTANT ATTRIBUTES	327
	<i>Principal Component Analysis</i>	328
	<i>Visual Techniques</i>	337
	<i>Kohonen Maps</i>	339
	<i>Workshop: PCA with R</i>	342
	<i>Further Reading</i>	348
15	INTERMEZZO: WHEN MORE IS DIFFERENT	351
	<i>A Horror Story</i>	353

<i>Some Suggestions</i>	354
<i>What About Map/Reduce?</i>	356
<i>Workshop: Generating Permutations</i>	357
<i>Further Reading</i>	358

PART IV Applications: Using Data

16	REPORTING, BUSINESS INTELLIGENCE, AND DASHBOARDS	361
	<i>Business Intelligence</i>	362
	<i>Corporate Metrics and Dashboards</i>	369
	<i>Data Quality Issues</i>	373
	<i>Workshop: Berkeley DB and SQLite</i>	376
	<i>Further Reading</i>	381
17	FINANCIAL CALCULATIONS AND MODELING	383
	<i>The Time Value of Money</i>	384
	<i>Uncertainty in Planning and Opportunity Costs</i>	391
	<i>Cost Concepts and Depreciation</i>	394
	<i>Should You Care?</i>	398
	<i>Is This All That Matters?</i>	399
	<i>Workshop: The Newsvendor Problem</i>	400
	<i>Further Reading</i>	403
18	PREDICTIVE ANALYTICS	405
	<i>Introduction</i>	405
	<i>Some Classification Terminology</i>	407
	<i>Algorithms for Classification</i>	408
	<i>The Process</i>	419
	<i>The Secret Sauce</i>	423
	<i>The Nature of Statistical Learning</i>	424
	<i>Workshop: Two Do-It-Yourself Classifiers</i>	426
	<i>Further Reading</i>	431
19	EPILOGUE: FACTS ARE NOT REALITY	433
A	PROGRAMMING ENVIRONMENTS FOR SCIENTIFIC COMPUTATION AND DATA ANALYSIS	435
	<i>Software Tools</i>	435
	<i>A Catalog of Scientific Software</i>	437
	<i>Writing Your Own</i>	443
	<i>Further Reading</i>	444
B	RESULTS FROM CALCULUS	447
	<i>Common Functions</i>	448
	<i>Calculus</i>	460
	<i>Useful Tricks</i>	468

<i>Notation and Basic Math</i>	472
<i>Where to Go from Here</i>	479
<i>Further Reading</i>	481
C WORKING WITH DATA	485
<i>Sources for Data</i>	485
<i>Cleaning and Conditioning</i>	487
<i>Sampling</i>	489
<i>Data File Formats</i>	490
<i>The Care and Feeding of Your Data Zoo</i>	492
<i>Skills</i>	493
<i>Terminology</i>	495
<i>Further Reading</i>	497
INDEX	499

Introduction

IMAGINE YOUR BOSS COMES TO YOU AND SAYS: “HERE ARE 50 GB OF LOGFILES—FIND A WAY TO IMPROVE OUR business!”

What would you do? Where would you start? And what would you do next?

It’s this kind of situation that the present book wants to help you with!

Data Analysis

Businesses sit on data, and every second that passes, they generate some more. Surely, there *must* be a way to make use of all this stuff. But how, exactly—that’s far from clear.

The task is difficult because it is so vague: there is no specific problem that needs to be solved. There is no specific question that needs to be answered. All you know is the overall *purpose*: improve the business. And all you have is “the data.” Where do you start?

You start with the only thing you have: “the data.” What is it? We don’t know! Although 50 GB sure sounds like a lot, we have no idea what it actually contains. The first thing, therefore, is to *take a look*.

And I mean this literally: the first thing to do is to *look* at the data by plotting it in different ways and looking at graphs. Looking at data, you will notice things—the way data points are distributed, or the manner in which one quantity varies with another, or the large number of outliers, or the total absence of them. . . . I don’t know what you will find, but there is no doubt: if you look at data, you will observe things!

These observations should lead to some reflection. “Ten percent of our customers drive ninety percent of our revenue.” “Whenever our sales volume doubles, the number of