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## R for Everyone

## Advanced Analytics and Graphics

Jared P. Lander

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## Foreword

R has had tremendous growth in popularity over the last three years. Based on that, you'd think that it was a new, up-and-coming language. But surprisingly, R has been around since 1993. Why the sudden uptick in popularity? The somewhat obvious answer seems to be the emergence of data science as a career and a field of study. But the underpinnings of data science have been around for many decades. Statistics, linear algebra, operations research, artificial intelligence, and machine learning all contribute parts to the tools that a modern data scientist uses. R, more than most languages, has been built to make most of these tools only a single function call away.

That's why I'm very excited to have this book as one of the first in the Addison-Wesley Data and Analytics Series. R is indispensable for many data science tasks. Many algorithms useful for prediction and analysis can be accessed through only a few lines of code, which makes it a great fit for solving modern data challenges. Data science as a field isn't just about math and statistics, and it isn't just about programming and infrastructure. This book provides a well-balanced introduction to the power and expressiveness of R and is aimed at a general audience.

I can't think of a better author to provide an introduction to R than Jared Lander. Jared and I first met through the New York City machine learning community in late 2009. Back then, the New York City data community was small enough to fit in a single conference room, and many of the other data meetups had yet to be formed. Over the last four years, Jared has been at the forefront of the emerging data science profession.

Through running the Open Statistical Programming Meetup, speaking at events, and teaching a course at Columbia on R, Jared has helped grow the community by educating programmers, data scientists, journalists, and statisticians alike. But Jared's expertise isn't limited to teaching. As an everyday practitioner, he puts these tools to use while consulting for clients big and small.

This book provides an introduction both to programming in R and to the various statistical methods and tools an everyday R programmer uses. Examples use publicly available datasets that Jared has helpfully cleaned and made accessible through his Web site. By using real data and setting up interesting problems, this book stays engaging to the end.

-Paul Dix, Series Editor

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## Preface

With the increasing prevalence of data in our daily lives, new and better tools are needed to analyze the deluge. Traditionally there have been two ends of the spectrum: lightweight, individual analysis using tools like Excel or SPSS and heavy duty, high-performance analysis built with C++ and the like. With the increasing strength of personal computers grew a middle ground that was both interactive and robust. Analysis done by an individual on his or her own computer in an exploratory fashion could quickly be transformed into something destined for a server, underpinning advanced business processes. This area is the domain of R, Python, and other scripted languages.

R, invented by Robert Gentleman and Ross Ihaka of the University of Auckland in 1993, grew out of S, which was invented by John Chambers at Bell Labs. It is a high-level language that was originally intended to be run interactively where the user runs a command, gets a result, and then runs another command. It has since evolved into a language that can also be embedded in systems and tackle complex problems.

In addition to transforming and analyzing data, R can produce amazing graphics and reports with ease. It is now being used as a full stack for data analysis, extracting and transforming data, fitting models, drawing inferences and making predictions, plotting and reporting results.

R's popularity has skyrocketed since the late 2000s, as it has stepped out of academia and into banking, marketing, pharmaceuticals, politics, genomics and many other fields. Its new users are often shifting from low-level, compiled languages like C++, other statistical packages such as SAS or SPSS, and from the 800-pound gorilla, Excel. This time period also saw a rapid surge in the number of add-on packages—libraries of prewritten code that extend R's functionality.

While R can sometimes be intimidating to beginners, especially for those without programming experience, I find that programming analysis, instead of pointing and clicking, soon becomes much easier, more convenient and more reliable. It is my goal to make that learning process easier and quicker.

This book lays out information in a way I wish I were taught when learning R in graduate school. Coming full circle, the content of this book was developed in conjuction with the data science course I teach at Columbia University. It is not meant to cover every minute detail of R, but rather the 20% of functionality needed to accomplish 80% of the work. The content is organized into self-contained chapters as follows.

Chapter 1, Getting R: Where to download R and how to install it. This deals with the varying operating systems and 32-bit versus 64-bit versions. It also gives advice on where to install R.

Chapter 2, The R Environment: An overview of using R, particularly from within RStudio. RStudio projects and Git integration are covered as is customizing and navigating RStudio.

Chapter 3, Packages: How to locate, install and load R packages.

Chapter 4, Basics of R: Using R for math. Variable types such as numeric, character and Date are detailed as are vectors. There is a brief introduction to calling functions and finding documentation on functions.

Chapter 5, Advanced Data Structures: The most powerful and commonly used data structure, data.frames, along with matrices and lists, are introduced.

Chapter 6, Reading Data into R: Before data can be analyzed it must be read into R. There are numerous ways to ingest data, including reading from CSVs and databases.

Chapter 7, Statistical Graphics: Graphics are a crucial part of preliminary data analysis and communicating results. R can make beautiful plots using its powerful plotting utilities. Base graphics and ggplot2 are introduced and detailed here.

Chapter 8, Writing R Functions: Repeatable analysis is often made easier with user-defined functions. The structure, arguments and return rules are discussed.

Chapter 9, Control Statements: Controlling the flow of programs using if, ifelse and complex checks.

Chapter 10, Loops, the Un-R Way to Iterate: Iterating using for and while loops. While these are generally discouraged they are important to know.

Chapter 11, Group Manipulation: A better alternative to loops, vectorization does not quite iterate through data so much as operate on all elements at once. This is more efficient and is primarily performed with the apply functions and plyr package.

Chapter 12, Data Reshaping: Combining multiple datasets, whether by stacking or joining, is commonly necessary as is changing the shape of data. The plyr and reshape2 packages offer good functions for accomplishing this in addition to base tools such as rbind, cbind and merge.

Chapter 13, Manipulating Strings: Most people do not associate character data with statistics but it is an important form of data. R provides numerous facilities for working with strings, including combining them and extracting information from within. Regular expressions are also detailed.

Chapter 14, Probability Distributions: A thorough look at the normal, binomial and Poisson distributions. The formulas and functions for many distributions are noted.

Chapter 15, Basic Statistics: These are the first statistics most people are taught, such as mean, standard deviation and t-tests.

Chapter 16, Linear Models: The most powerful and common tool in statistics, linear models are extensively detailed.

Chapter 17, Generalized Linear Models: Linear models are extended to include logistic and Poisson regression. Survival analysis is also covered.

Chapter 18, Model Diagnostics: Determining the quality of models and variable selection using residuals, AIC, cross-validation, the bootstrap and stepwise variable selection.

Chapter 19, Regularization and Shrinkage: Preventing overfitting using the Elastic Net and Bayesian methods.

Chapter 20, Nonlinear Models: When linear models are inappropriate, nonlinear models are a good solution. Nonlinear least squares, splines, generalized additive models, decision trees and random forests are discussed.

Chapter 21, Time Series and Autocorrelation: Methods for the analysis of univariate and multivariate time series data.

Chapter 22, Clustering: Clustering, the grouping of data, is accomplished by various methods such as K-means and hierarchical clustering.

Chapter 24, Building R Packages: R packages are great for portable, reusable code. Building these packages has been made incredibly easy with the advent of devtools and Rcpp.

Appendix A, Real-Life Resources: A listing of our favorite resources for learning more about R and interacting with the community.

Appendix B, Glossary: A glossary of terms used throughout this book.

A good deal of the text in this book is either R code or the results of running code. Code and results are most often in a separate block of text and set in a distinctive font, as shown in the following example. The different parts of code also have different colors. Lines of code start with >, and if code is continued from one line to another the continued line begins with +.

```
> # this is a comment
>
> # now basic math
> 10 * 10
[1] 100
>
> # calling a function
> sqrt(4)
```

#### [1] 2

Certain Kindle devices do not display color so the digital edition of this book will be viewed in greyscale on those devices.

There are occasions where code is shown inline and looks like sqrt(4).

In the few places where math is necessary, the equations are indented from the margin and are numbered.

$$e^{i\pi} + 1 = 0 \tag{1}$$

Within equations, normal variables appear as italic text (x), vectors are bold lowercase letters (x) and matrices are bold uppercase letters (X). Greek letters, such as  $\alpha$  and  $\beta$ , follow the same convention.

Function names will be written as join and package names as plyr. Objects generated in code that are referenced in text are written as object1.

Learning R is a gratifying experience that makes life so much easier for so many tasks. I hope you enjoy learning with me.

## Acknowledgments

To start, I must thank my mother, Gail Lander, for encouraging me to become a math major. Without that I would never have followed the path that led me to statistics and data science. In a similar vein, I have to thank my father, Howard Lander, for paying all those tuition bills. He has been a valuable source of advice and guidance throughout my life and someone I have aspired to emulate in many ways. While they both insist they do not understand what I do, they love that I do it and have helped me all along the way. Staying with family, I should thank my sister and brother-in-law, Aimee and Eric Schechterman, for letting me teach math to Noah, their five-year-old son.

There are many teachers who have helped shape me over the years. The first is Rochelle Lecke, who tutored me in middle school math even when my teacher told me I did not have worthwhile math skills.

Then there is Beth Edmondson, my precalc teacher at Princeton Day School. After I wasted the first half of high school as a mediocre student, she told me I had "some nerve signing up for next year's AP Calc" given my grades. She agreed to let me take AP Calc if I went from a C to an A+ in her class, never thinking I stood a chance. Three months later, she was in shock as I not only earned the A+, but turned around my entire academic career. She changed my life and without her, I do not know where I would be today. I am forever grateful that she was my teacher.

For the first two years at Muhlenberg College, I was determined to be a business and communications major, but took math classes because they came naturally to me. My professors, Dr. Penny Dunham, Dr. Bill Dunham, and Dr. Linda McGuire, all convinced me to become a math major, a decision that has greatly shaped my life. Dr. Greg Cicconetti gave me my first glimpse of rigorous statistics, my first research opportunity and planted the idea in my head that I should go to grad school for statistics.

While earning my M.A. at Columbia University, I was surrounded by brilliant minds in statistics and programming. Dr. David Madigan opened my eyes to modern machine learning, and Dr. Bodhi Sen got me thinking about statistical programming. I had the privilege to do research with Dr. Andrew Gelman, whose insights have been immeasurably important to me. Dr. Richard Garfield showed me how to use statistics to help people in disaster and war zones when he sent me on my first assignment to Myanmar. His advice and friendship over the years have been dear to me. Dr. Jingchen Liu ΧХ

allowed and encouraged me to write my thesis on New York City pizza, which has brought me an inordinate amount of attention.<sup>1</sup>

While at Columbia, I also met my good friend—and one time TA— Dr. Ivor Cribben who filled in so many gaps in my knowledge. Through him, I met Dr. Rachel Schutt, a source of great advice, and who I am now honored to teach alongside at Columbia.

Grad school might never have happened without the encouragement and support of Shanna Lee. She helped maintain my sanity while I was incredibly overcommited to two jobs, classes and Columbia's hockey team. I am not sure I would have made it through without her.

Steve Czetty gave me my first job in analytics at Sky IT Group and taught me about databases, while letting me experiment with off-the-wall programming. This sparked my interest in statistics and data. Joe DeSiena, Philip du Plessis, and Ed Bobrin at the Bardess Group are some of the finest people I have ever had the pleasure to work with, and I am proud to be working with them to this day. Mike Minelli, Rich Kittler, Mark Barry, David Smith, Joseph Rickert, Dr. Norman Nie, James Peruvankal, Neera Talbert and Dave Rich at Revolution Analytics let me do one of the best jobs I could possibly imagine: explaining to people in business why they should be using R. Kirk Mettler, Richard Schultz, Dr. Bryan Lewis and Jim Winfield at Big Computing encouraged me to have fun, tackling interesting problems in R. Vincent Saulys, John Weir, and Dr. Saar Golde at Goldman Sachs made my time there both enjoyable and educational.

Throughout the course of writing this book, many people helped me with the process. First and foremost is Yin Cheung, who saw all the stress I constantly felt and supported me through many ruined nights and days.

My editor, Debra Williams, knew just how to encourage me and her guiding hand has been invaluable. Paul Dix, the series editor and a good friend, was the person who suggested I write this book, so none of this would have happened without him. Thanks to Caroline Senay and Andrea Fox for being great copy editors. Without them, this book would not be nearly as well put together. Robert Mauriello's technical review was incredibly useful in honing the book's presentation.

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Numerous people have provided valuable feedback as I produced this book, including Chris Bethel, Dr. Dirk Eddelbuettel, Dr. Ramnath Vaidyanathan, Dr. Eran Bellin,

<sup>1.</sup> http://slice.seriouseats.com/archives/2010/03/the-moneyball-of-pizza-

statistician-uses-statistics-to-find-nyc-best-pizza.html

Avi Fisher, Brian Ezra, Paul Puglia, Nicholas Galasinao, Aaron Schumaker, Adam Hogan, Jeffrey Arnold, and John Houston.

Last fall was my first time teaching, and I am thankful to the students from the Fall 2012 Introduction to Data Science class at Columbia University for being the guinea pigs for the material that ultimately ended up in this book.

Thank you to everyone who helped along the way.

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## About the Author

**Jared P. Lander** is the founder and CEO of Lander Analytics, a statistical consulting firm based in New York City, the organizer of the New York Open Statistical Programming Meetup, and an adjunct professor of statistics at Columbia University. He is also a tour guide for Scott's Pizza Tours and an advisor to Brewla Bars, a gourmet ice pop start-up. With an M.A. from Columbia University in statistics and a B.A. from Muhlenberg College in mathematics, he has experience in both academic research and industry. His work for both large and small organizations spans politics, tech start-ups, fund-raising, music, finance, healthcare and humanitarian relief efforts.

He specializes in data management, multilevel models, machine learning, generalized linear models, visualization, data management and statistical computing.

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## Chapter 12 Data Reshaping

As noted in Chapter 11, manipulating the data takes a great deal of effort before serious analysis can begin. In this chapter we will consider when the data needs to be rearranged from column oriented to row oriented (or the opposite) and when the data are in multiple, separate sets and need to be combined into one.

There are base functions to accomplish these tasks but we will focus on those in plyr, reshape2 and data.table.

#### 12.1 cbind and rbind

The simplest case is when we have two datasets with either identical columns (both the number of and names) or the same number of rows. In this case, either rbind or cbind work great.

As a first trivial example, we create two simple data.frames by combining a few vectors with cbind, and then stack them using rbind.

```
> # make two vectors and combine them as columns in a data.frame
> sport <- c("Hockey", "Baseball", "Football")</pre>
> league <- c("NHL", "MLB", "NFL")</pre>
> trophy <- c("Stanley Cup", "Commissioner's Trophy",</pre>
               "Vince Lombardi Trophy")
+
> trophies1 <- cbind(sport, league, trophy)</pre>
> # make another data.frame using data.frame()
> trophies2 <- data.frame(sport=c("Basketball", "Golf"),</pre>
+
                            league=c("NBA", "PGA"),
                            trophy=c("Larry O'Brien Championship Trophy",
+
                                      "Wanamaker Trophy"),
+
                            stringsAsFactors=FALSE)
> # combine them into one data.frame with rbind
> trophies <- rbind(trophies1, trophies2)</pre>
```

Both cbind and rbind can take multiple arguments to combine an arbitrary number of objects. Note that it is possible to assign new column names to vectors in cbind.

```
> cbind(Sport = sport, Association = league, Prize = trophy)
        Sport Association Prize
[1,] "Hockey" "NHL" "Stanley Cup"
[2,] "Baseball" "MLB" "Commissioner's Trophy"
[3,] "Football" "NFL" "Vince Lombardi Trophy"
```

#### 12.2 Joins

Data do not always come so nicely aligned for combining using cbind, so they need to be joined together using a common key. This concept should be familiar to SQL users. Joins in R are not as flexible as SQL joins, but are still an essential operation in the data analysis process.

The three most commonly used functions for joins are merge in base R, join in plyr and the merging functionality in data.table. Each has pros and cons with some pros outweighing their respective cons.

To illustrate these functions I have prepared data originally made available as part of the USAID Open Government initiative.<sup>1</sup> The data have been chopped into eight separate files so that they can be joined together. They are all available in a zip file at http://jaredlander.com/data/US\_Foreign\_Aid.zip. These should be downloaded and unzipped to a folder on our computer. This can be done a number of ways (including using a mouse!) but we show how to download and unzip using R.

To load all of these files programmatically, we use a for loop as seen in Section 10.1. We get a list of the files using dir, and then loop through that list assigning each dataset to a name specified using assign.

<sup>1.</sup> More information about the data is available at http://gbk.eads.usaidallnet.gov/.

#### 12.2.1 merge

R comes with a built-in function, called merge, to merge two data.frames.

```
> Aid90s00s <- merge(x=Aid_90s, y=Aid_00s,
+ by.x=c("Country.Name", "Program.Name"),
+ by.y=c("Country.Name", "Program.Name"))
> head(Aid90s00s)
```

	Country.Name Program.Name								
1	Afghar	nistan	Child Survival and Health						
2	Afghar	nistan	an Department of Defense Security Assistance						
3	Afghar	nistan				Deve	lopment	Assista	ance
4	Afghar	nistan	Economic	Support	Fund/Se	curity	Support	Assista	ance
5								cion	
6	Afghar	nistan			Global	Health	and Chil	ld Surv	ival
	FY1990	FY199	1 FY1992	FY1993	FY1994	FY1995	FY1996	FY1997	FY1998
1	NA	N	A NA	NA	NA	NA	NA	NA	NA
2	NA	N	A NA	NA	NA	NA	NA	NA	NA
3	NA	N	A NA	NA	NA	NA	NA	NA	NA
4	NA	N	A NA	14178135	2769948	NA	NA	NA	NA
5	NA	N	A NA	NA	NA	NA	NA	NA	NA
6	NA	N	A NA	NA	NA	NA	NA	NA	NA
	FY1999	FY200	) FY2001	FY2002	FY2	003		FY2004	FY2005
1	NA	N	A NA	2586555	56501	189	40	0215304	39817970
2	NA	N	A NA	2964313		NA	45	5635526	151334908
3	NA	N	A 4110478	8762080	54538	965	180	)539337	193598227
4	NA	N	A 61144	31827014	341306	822	1025	5522037	1157530168
5	NA	N	A NA	. NA	3957	312	2	2610006	3254408
6	NA	N	A NA	. NA	L	NA		NA	NA
	FY2	2006	FY2007	FY20	08	FY2009			
1	40856	382	72527069	283974	35	NA			
2	230501	318	214505892	4955390	84 552	524990			
3	212648	3440	173134034	1505298	62 3	675202			
4	1357750	249 1	266653993	14002377	91 1418	688520			
5	386	891	NA		NA	NA			
6		NA	NA	630649	12 1	764252			

The by.x specifies the key column(s) in the left data.frame and by.y does the same for the right data.frame. The ability to specify different column names for each data.frame is the most useful feature of merge. The biggest drawback, however, is that merge can be much slower than the alternatives.

#### 12.2.2 plyr join

Returning to Hadley Wickham's plyr package, we see it includes a join function, which works similarly to merge but is much faster. The biggest drawback, though, is that the key column(s) in each table must have the same name. We use the same data used previously to illustrate.

```
> require(plyr)
> Aid90s00sJoin <- join(x = Aid_90s, y = Aid_00s, by = c("Country.Name",
      "Program.Name"))
+
> head(Aid90s00sJoin)
  Country.Name
                                                         Program.Name
1
   Afghanistan
                                           Child Survival and Health
2
  Afghanistan
                         Department of Defense Security Assistance
3
   Afghanistan
                                              Development Assistance
4
  Afghanistan Economic Support Fund/Security Support Assistance
5
   Afghanistan
                                                  Food For Education
   Afghanistan
6
                                   Global Health and Child Survival
  FY1990 FY1991 FY1992
                           FY1993
                                    FY1994 FY1995 FY1996 FY1997 FY1998
1
      NA
              NA
                      NA
                               NA
                                        NA
                                                NA
                                                        NA
                                                               NA
                                                                       NΑ
2
      NA
                      NA
                               NA
                                        NA
                                                NA
                                                                       NA
              NΑ
                                                        NA
                                                               NΑ
3
      NA
                               NA
                                        NA
                                                NA
              NΑ
                      NA
                                                        NA
                                                               NA
                                                                       NA
4
      NA
              NA
                      NA 14178135 2769948
                                                NA
                                                               NA
                                                                       NA
                                                        NΑ
5
      NA
                      NA
                               NA
                                        NA
                                                NA
                                                        NA
                                                               NA
                                                                       NA
              NΑ
6
      NA
                               NA
                                        NA
                                                NA
                                                                       NA
              NΑ
                      NΑ
                                                        NΑ
                                                               NA
  FY1999 FY2000
                  FY2001
                            FY2002
                                       FY2003
                                                   FY2004
                                                               FY2005
1
      NΑ
              NΑ
                       NΑ
                           2586555
                                     56501189
                                                 40215304
                                                             39817970
                           2964313
2
      NΑ
              NΑ
                       NΑ
                                            NΑ
                                                 45635526
                                                            151334908
3
      NA
              NA 4110478
                           8762080
                                     54538965
                                                180539337
                                                            193598227
4
      NA
              NA
                   61144 31827014 341306822 1025522037 1157530168
5
                                      3957312
                                                  2610006
                                                              3254408
      NA
              NA
                       NA
                                 NA
6
                                                        NA
                                                                    NA
      NA
                       NA
                                 NA
                                           NA
              NA
      FY2006
                                           FY2009
                  FY2007
                              FY2008
                72527069
1
    40856382
                            28397435
                                               NA
2
   230501318
               214505892
                           495539084
                                       552524990
3
   212648440
               173134034
                           150529862
                                         3675202
4
  1357750249 1266653993 1400237791 1418688520
5
      386891
                       NA
                                   NA
                                               NA
6
                            63064912
           NA
                       NA
                                         1764252
```

join has an argument for specifying a left, right, inner or full (outer) join.

We have eight data.frames containing foreign assistance data that we would like to combine into one data.frame without hand coding each join. The best way to do this is to put all the data.frames into a list, and then successively join them together using Reduce.

```
> # first figure out the names of the data.frames
> frameNames <- str_sub(string = theFiles, start = 12, end = 18)
> # build an empty list
> frameList <- vector("list", length(frameNames))
> names(frameList) <- frameNames
> # add each data.frame into the list
> for (a in frameNames)
+ {
+ frameList[[a]] <- eval(parse(text = a))
+ }
```

A lot happened in that section of code, so let's go over it carefully. First we reconstructed the names of the data.frames using str\_sub from Hadley Wickham's stringr package, which is shown in more detail in Chapter 13. Then we built an empty list with as many elements as there are data.frames, in this case eight, using vector and assigning its mode to "list." We then set appropriate names to the list.

Now that the list is built and named, we loop through it, assigning to each element the appropriate data.frame. The problem is that we have the names of the data.frames as characters but the <- operator requires a variable, not a character. So we parse and evaluate the character, which realizes the actual variable. Inspecting, we see that the list does indeed contain the appropriate data.frames.

```
> head(frameList[[1]])
```

	Country.Name Program.Name										
1	Afghar	nanistan Child Survival and Health									
2	Afghar	nistan	Dep	Department of Defense Security Assistance							
3	Afghanistan Development Assistance						istance				
4	4 Afghanistan Economic Support Fund/Security Support Assistance										
5	5 Afghanistan Food For Education										
6	Afghar	Afghanistan Global Health and Child Survival									
	FY2000	FY2001	FY2002	FY2003	FY2004	FY2005	FY2006				
1	NA	NA	JA 2586555 56501189 40215304 39817970 40856								
2	NA	A NA 2964313 NA 45635526 45635526 23050131									
3	NA	NA 4110478 8762080 54538965 180539337 193598227 2126484									
4	NA	61144	31827014	341306822	1025522037	1157530168	1357750249				
5	NA	NA	NA	3957312	2610006	3254408	386891				
6	NA	NA	NA	NA	NA	NA	NA				

NA

NA

NA

NA

NA

NA

NA

NA

NA

FY2007 FY2008 FY2009 1 72527069 28397435 NA 2 214505892 495539084 552524990 3 173134034 150529862 3675202 4 1266653993 1400237791 1418688520 5 NA NA NA 6 NA 63064912 1764252 > head(frameList[["Aid\_00s"]]) Country.Name Program.Name 1 Afghanistan Child Survival and Health 2 Afghanistan Department of Defense Security Assistance 3 Afghanistan Development Assistance 4 Afghanistan Economic Support Fund/Security Support Assistance 5 Afghanistan Food For Education Afghanistan Global Health and Child Survival 6 FY2000 FY2001 FY2003 FY2004 FY2005 FY2006 FY2002 1 NA NA 2586555 56501189 40215304 39817970 40856382 2 NA NA 2964313 45635526 151334908 230501318 NA 3 NA 4110478 8762080 54538965 180539337 193598227 212648440 61144 31827014 341306822 1025522037 1157530168 1357750249 4 NA 5 3254408 NA NA NA 3957312 2610006 386891 6 NA NA NA NA NA NA NA FY2008 FY2007 FY2009 72527069 28397435 1 NA 2 214505892 495539084 552524990 3 173134034 150529862 3675202 4 1266653993 1400237791 1418688520 5 NA NA NA 6 NA 63064912 1764252 > head(frameList[[5]]) Country.Name Program.Name 1 Afghanistan Child Survival and Health 2 Afghanistan Department of Defense Security Assistance 3 Afghanistan Development Assistance 4 Afghanistan Economic Support Fund/Security Support Assistance 5 Afghanistan Food For Education 6 Afghanistan Global Health and Child Survival FY1960 FY1961 FY1962 FY1963 FY1964 FY1965 FY1966 FY1967 FY1968 1 NA NA NA NA NA NA NA NA NA 2 NA NA NA NA NA NA NA NA NA 3

4	NA	NA	181177853	NA	NA	NA	NA	NA	NA
5	NA	NA	NA	NA	NA	NA	NA	NA	NA
6	NA	NA	NA	NA	NA	NA	NA	NA	NA
	FY1969								
1	NA								
2	NA								
3	NA								
4	NA								
5	NA								
6	NA								
>	<pre>&gt; head(frameList[["Aid_60s"]])</pre>								
	Country							rogram.I	
1	Afghar						urvival		
2	Afghar		Dep	partment	t of De		ecurity		
3	Afghar						lopment		
4		Afghanistan Economic Support Fund/Security Support Assistance							ance
	5 Afghanistan Food For Education								
5	-								
5 6	Afghar	nistan				Health a	and Chil	Ld Surv:	lval
6	Afghar FY1960	nistan FY1961		FY1963	FY1964	Health a FY1965	and Chil FY1966	ld Surv: FY1967	ival FY1968
6 1	Afghar FY1960 NA	nistan FY1961 NA	NA	FY1963 NA	FY1964 NA	Health a FY1965 NA	and Chi FY1966 NA	ld Surv: FY1967 NA	ival FY1968 NA
6 1 2	Afghar FY1960 NA NA	nistan FY1961 NA NA	NA	FY1963 NA NA	FY1964 NA NA	Health a FY1965 NA NA	and Chil FY1966 NA NA	ld Surv: FY1967 NA NA	ival FY1968 NA NA
6 1 2 3	Afghar FY1960 NA NA NA	nistan FY1961 NA NA NA	NA NA NA	FY1963 NA NA NA	FY1964 NA NA NA	Health a FY1965 NA NA NA	and Chil FY1966 NA NA NA	ld Surv: FY1967 NA NA NA	ival FY1968 NA NA NA
6 1 2 3 4	Afghar FY1960 NA NA NA	FY1961 NA NA NA NA	NA NA NA 181177853	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA	and Chi FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	IVAL FY1968 NA NA NA
6 1 2 3 4 5	Afghar FY1960 NA NA NA NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4	Afghar FY1960 NA NA NA NA NA	FY1961 NA NA NA NA	NA NA NA 181177853	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA	and Chi FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	IVAL FY1968 NA NA NA
6 1 2 3 4 5 6	Afghar FY1960 NA NA NA NA NA FY1969	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4 5 6 1	Afghar FY1960 NA NA NA NA FY1969 NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4 5 6 1 2	Afghar FY1960 NA NA NA NA FY1969 NA NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4 5 6 1 2 3	Afghar FY1960 NA NA NA NA FY1969 NA NA NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4 5 6 1 2 3 4	Afghar FY1960 NA NA NA NA FY1969 NA NA NA NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA
6 1 2 3 4 5 6 1 2 3	Afghar FY1960 NA NA NA NA FY1969 NA NA NA	histan FY1961 NA NA NA NA	NA NA 181177853 NA	FY1963 NA NA NA NA	FY1964 NA NA NA NA	Health a FY1965 NA NA NA NA NA	and Chil FY1966 NA NA NA NA	ld Surv: FY1967 NA NA NA NA	ival FY1968 NA NA NA NA

Having all the data.frames in a list allows us to iterate through the list, joining all the elements together (or applying any function to the elements iteratively). Rather than using a loop, we use the Reduce function to speed up the operation.

```
> allAid <- Reduce(function(...)
+ {
+     join(..., by = c("Country.Name", "Program.Name"))
+ }, frameList)
> dim(allAid)
[1] 2453 67
```

```
> require(useful)
> corner(allAid, c = 15)
  Country.Name
                                                     Program.Name
  Afghanistan
                                        Child Survival and Health
1
  Afghanistan
                       Department of Defense Security Assistance
2
3 Afghanistan
                                           Development Assistance
4
  Afghanistan Economic Support Fund/Security Support Assistance
5
  Afghanistan
                                               Food For Education
  FY2000 FY2001
                   FY2002
                             FY2003
                                         FY2004
                                                    FY2005
                                                               FY2006
1
      NA
              NA 2586555 56501189
                                       40215304
                                                  39817970
                                                             40856382
2
      NA
              NA
                 2964313
                                 NA
                                      45635526 151334908 230501318
3
      NA 4110478
                  8762080 54538965
                                     180539337 193598227
                                                            212648440
4
      NA
           61144 31827014 341306822 1025522037 1157530168 1357750249
5
                            3957312
                                        2610006
                                                   3254408
                                                               386891
      NA
              NA
                       NA
                            FY2009
                                        FY2010 FY1946 FY1947
      FY2007
                 FY2008
1
    72527069
               28397435
                                NA
                                           NA
                                                    NA
                                                           NA
2
  214505892 495539084 552524990
                                    316514796
                                                           NA
                                                    NA
3
  173134034 150529862
                           3675202
                                                           NA
                                            NA
                                                    NA
4 1266653993 1400237791 1418688520 2797488331
                                                           NA
                                                    NA
5
          NA
                     NA
                                NA
                                            NA
                                                    NA
                                                           NA
```

> bottomleft(allAid, c = 15)

	Country.1	Name	Pro	gram.Name	FY2000	FY2001	FY2002
2449	Zimba	abwe Othe	r State A	State Assistance		322842	NA
2450	Zimba	abwe Othe	r USAID A	ssistance	3033599	8464897 6	624408
2451	Zimba	abwe	Pe	ace Corps	2140530	1150732	407834
2452	Zimba	abwe		Title I	NA	NA	NA
2453	Zimba	abwe		Title II		NA 31	019776
	FY2003	FY2004	FY2005	FY2006	FY2007	FY2008	FY2009
2449	NA	318655	44553	883546	1164632	2455592	2193057
2450	11580999	12805688	10091759	4567577	10627613	11466426	41940500
2451	NA	NA	NA	NA	NA	NA	NA
2452	NA	NA	NA	NA	NA	NA	NA
2453	NA	NA	NA	277468	100053600	180000717	174572685
	FY2010	FY1946 F	Y1947				
2449	1605765	NA	NA				
2450	30011970	NA	NA				
2451	NA	NA	NA				
2452	NA	NA	NA				
2453	79545100	NA	NA				

Reduce can be a difficult function to grasp, so we illustrate it with a simple example. Let's say we have a vector of the first ten integers, 1:10, and want to sum them (forget for a moment that sum(1:10) will work perfectly). We can call Reduce(sum, 1:10),

which will first add 1 and 2. It will then add 3 to that result, then 4 to that result, and so on, resulting in 55.

Likewise, we passed a list to a function that joins its inputs, which in this case was simply ..., meaning that anything could be passed. Using ... is an advanced trick of R programming that can be difficult to get right. Reduce passed the first two data.frames in the list, which were then joined. That result was then joined to the next data.frame and so on until they were all joined together.

#### 12.2.3 data.table merge

Like many other operations in data.table, joining data requires a different syntax, and possibly a different way of thinking. To start, we convert two of our foreign aid datasets' data.frames into data.tables.

```
> require(data.table)
> dt90 <- data.table(Aid_90s, key = c("Country.Name", "Program.Name"))
> dt00 <- data.table(Aid_00s, key = c("Country.Name", "Program.Name"))</pre>
```

Then, doing the join is a simple operation. Note that the join requires specifying the keys for the data.tables, which we did during their creation.

```
> dt0090 <- dt90[dt00]
```

In this case dt90 is the left side, dt00 is the right side and a left join was performed.

#### 12.3 reshape2

The next most common munging need is either melting data (going from column orientation to row orientation) or casting data (going from row orientation to column orientation). As with most other procedures in R, there are multiple functions available to accomplish these tasks but we will focus on Hadley Wickham's reshape2 package. (We talk about Wickham a lot because his products have become so fundamental to the R developer's toolbox.)

#### 12.3.1 melt

Looking at the Aid\_00s data.frame, we see that each year is stored in its own column. That is, the dollar amount for a given country and program is found in a different column for each year. This is called a cross table, which, while nice for human consumption, is not ideal for graphing with ggplot2 or for some analysis algorithms.

```
> head(Aid_00s)
```

	Country.Name				Pi	rogra	am.Name
1	Afghanistan			Child	Survival	and	Health
2	Afghanistan	Department	of	Defense	Security	Assi	istance

3	3 Afghanistan Development Assistance										
4	4 Afghanistan Economic Support Fund/Security Support Assistance										
5	5 Afghanistan Food For Education										
6	6 Afghanistan Global Health and Child Survival										
	FY2000	FY200	1 FY20	)2 FY	2003	FY2004	FY2005	FY2006			
1	NA	N	IA 258655	55 5650	1189	40215304	39817970	40856382			
2	NA	N	IA 296433	13	NA	45635526	151334908	230501318			
3	NA	411047	8 876208	30 5453	8965	180539337	193598227	212648440			
4	NA	6114	4 3182703	14 34130	6822	1025522037	1157530168	1357750249			
5	NA	N	IA I	NA 395	7312	2610006	3254408	386891			
6	NA	N	IA I	A	NA	NA	NA	NA			
	FY2	007	FY2008	FY2	009						
1	72527	069	28397435		NA						
2	214505	892 4	95539084	552524	990						
3	173134	034 1	50529862	3675	202						
4	1266653	993 14	00237791	1418688	520						
5		NA	NA		NA						
6		NA	63064912	1764	252						

We want it set up so that each row represents a single country-program-year entry with the dollar amount stored in one column. To achieve this we melt the data using melt from reshape2.

```
> require(reshape2)
> melt00 <- melt(Aid_00s, id.vars=c("Country.Name", "Program.Name"),</pre>
                 variable.name="Year", value.name="Dollars")
+
> tail(melt00, 10)
      Country.Name
24521
          7.imbabwe
24522
          Zimbabwe
24523
          Zimbabwe
24524
          Zimbabwe
24525
          Zimbabwe
          Zimbabwe
24526
24527
          Zimbabwe
          Zimbabwe
24528
24529
          Zimbabwe
24530
          Zimbabwe
                                                  Program.Name
                                                                  Year
24521
                             Migration and Refugee Assistance FY2009
24522
                                             Narcotics Control FY2009
```

24523	Nonproliferation,	Anti-Terrorism, Demining and Related	FY2009
24524		Other Active Grant Programs	FY2009
24525		Other Food Aid Programs	FY2009
24526		Other State Assistance	FY2009
24527		Other USAID Assistance	FY2009
24528		Peace Corps	FY2009
24529		Title I	FY2009
24530		Title II	FY2009
	Dollars		
24521	3627384		
24522	NA		
24523	NA		
24524	7951032		
24525	NA		
24526	2193057		
24527	41940500		
24528	NA		
24529	NA		
24530	174572685		

The id.vars argument specifies which columns uniquely identify a row.

After some manipulation of the Year column and aggregating, this is now prime for plotting, as shown in Figure 12.1. The plot uses faceting allowing us to quickly see and understand the funding for each program over time.

```
> require(scales)
> # strip the "FY" out of the year column and convert it to numeric
> melt00$Year <- as.numeric(str_sub(melt00$Year, start=3, 6))</pre>
> # aggregate the data so we have yearly numbers by program
> meltAgg <- aggregate(Dollars ~ Program.Name + Year, data=melt00,
+
                        sum, na.rm=TRUE)
> # just keep the first 10 characters of program name
> # then it will fit in the plot
> meltAgg$Program.Name <- str_sub(meltAgg$Program.Name, start=1,</pre>
+
                                   end=10)
>
> ggplot(meltAgg, aes(x=Year, y=Dollars)) +
      geom_line(aes(group=Program.Name)) +
^{+}
      facet_wrap(~ Program.Name) +
+
+
      scale x continuous(breaks=seq(from=2000, to=2009, by=2)) +
      theme(axis.text.x=element_text(angle=90, vjust=1, hjust=0)) +
+
      scale_y_continuous(labels=multiple_format(extra=dollar,
^{+}
+
                                                  multiple="B"))
```

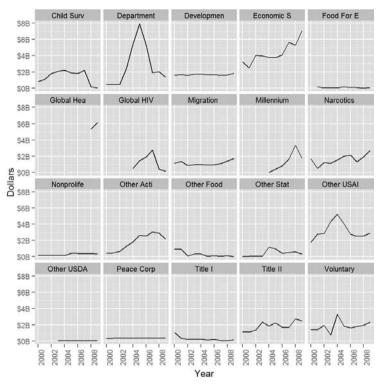


Figure 12.1 Plot of foreign assistance by year for each of the programs.

## 12.3.2 dcast

Now that we have the foreign aid data melted, we cast it back into the wide format for illustration purposes. The function for this is dcast, and it has trickier arguments than melt. The first is the data to be used, in our case melt00. The second argument is a formula where the left side specifies the columns that should remain columns and the right side specifies the columns that should become row names. The third argument is the column (as a character) that holds the values to be populated into the new columns representing the unique values of the right side of the formula argument.

```
> cast00 <- dcast(melt00, Country.Name + Program.Name ~ Year,</pre>
      value.var = "Dollars")
+
> head(cast00)
  Country.Name
                                                       Program.Name 2000
  Afghanistan
                                         Child Survival and Health
1
                                                                       NA
2
  Afghanistan
                        Department of Defense Security Assistance
                                                                       NA
3
   Afghanistan
                                            Development Assistance
                                                                       NA
```

```
Afghanistan Economic Support Fund/Security Support Assistance
                                                                        NA
4
5
   Afghanistan
                                                  Food For Education
                                                                        NA
6
  Afghanistan
                                  Global Health and Child Survival
                                                                        NA
     2001
               2002
                          2003
                                      2004
                                                  2005
                                                             2006
1
       NA
           2586555
                     56501189
                                 40215304
                                             39817970
                                                         40856382
       NA
2
           2964313
                            NA
                                 45635526
                                            151334908
                                                        230501318
3
  4110478
           8762080
                     54538965
                                180539337
                                            193598227
                                                        212648440
    61144 31827014 341306822 1025522037 1157530168 1357750249
4
5
                                  2610006
                                              3254408
                                                           386891
       NA
                 NA
                       3957312
6
       NA
                 NA
                            NA
                                        NA
                                                    NA
                                                                NA
        2007
                    2008
                                2009
1
    72527069
                28397435
                                  NA
2
   214505892
               495539084
                           552524990
3
   173134034
               150529862
                             3675202
4
  1266653993 1400237791 1418688520
5
          NA
                      NA
                                  NA
                63064912
6
          NA
                             1764252
```

## **12.4** Conclusion

Getting the data just right to analyze can be a time-consuming part of our work flow, although it is often inescapable. In this chapter we examined combining multiple datasets into one and changing the orientation from column based (wide) to row based (long). We used plyr, reshape2 and data.table along with base functions to accomplish this. This chapter combined with Chapter 11 covers most of the basics of data munging with an eye to both convenience and speed. This page intentionally left blank

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