



Stat 107: Introduction to Business and Financial Statistics Class 3: Computing Discussion

### | Introduction to R



# Why R?

- Free (open-source)
- Programming language (not point-and-click)
- **Excellent graphics**
- Offers broadest range of statistical tools
- Easy to generate reproducible reports
- Easy to integrate with other tools
- What the cool kids are using
- Used in 110,111,139,121, etc.......

#### R is a tool for...

#### Data Manipulation

- connecting to data sources
- slicing & dicing data

#### Modeling & Computation

- statistical modeling
- numerical simulation

#### Data Visualization

- visualizing fit of models
- composing statistical graphics

# munge model visualize

#### The R Console

- Basic interaction with R is through typing in the console
- This is the terminal or command-line interface



### Rstudio is an IDE for R

- RStudio has 4 main windows ('panes'):
  - Source
  - Console
  - Workspace/History
  - ☐ Files/Plots/Packages/Help



### History of R

- S: language for data analysis developed at Bell Labs circa 1976
- Licensed by AT&T/Lucent to Insightful Corp. Product name: S-plus.
- R: initially written & released as an open source software by Ross Ihaka and Robert Gentleman at U Auckland during 90s (R plays on name "S")
- Since 1997: international R-core team ~15 people & 1000s of code writers and statisticians happy to share their libraries! AWESOME!

### "Open source"... that just means I don't have to pay for it, right?

•No. Much more:

- -Provides full access to algorithms and their implementation
- -Gives you the ability to fix bugs and extend software
- -Provides a forum allowing researchers to explore and expand the methods used to analyze data
- -Is the product of 1000s of leading experts in the fields they know best. It is CUTTING EDGE
- -Ensures that scientists around the world and not just ones in rich countries are the co-owners to the software tools needed to carry out research
- -Promotes reproducible research by providing open and accessible tools
- -Most of R is written in... R! This makes it quite easy to see what functions are actually doing.

### There are over 800 add-on packages

- This is an enormous advantage new techniques available without delay, and they can be performed using the R language you already know.
- Allows you to build a customized statistical program suited to your own needs.
- Downside = as the number of packages grows, it is becoming difficult to choose the best package for your needs, & QC is an issue.

(http://cran.r-project.org/src/contrib/PACKAGES.html)

#### http://cran.r-project.org/web/views/Finance.html

- The Rmetrics suite of packages comprises [Arma, [AsianOptios [Noulinear, (Options, (Portfolio, (Repression, timeSeries) (formes motion-grantland finance ovides several option-prising function. The computational finance ovides several option-prising function of the quantimed package offers a number of financions for quantiti. The particles package contains classes for equity portfolio manumarket prices. The backets offers tools to explore portfolio-ba multigroup models.

  The Performance Cambrides package contains a large number of:
  The TITE contains functions to construct technical trading rules i The Fannacial package can be properly the properly of the trade of the trade

Risk management

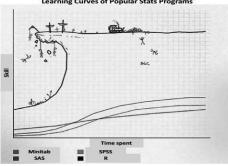
### Again- Why R?

- It runs on a variety of platforms including Windows, Unix and
- It provides an unparalleled platform for programming new statistical methods in an easy and straightforward manner.
- It contains advanced statistical routines not yet available in other packages.
- It has state-of-the-art graphics capabilities.

### Warning: R has a (steep) learning curve

- First, while there are many introductory tutorials (covering data types, basic commands, the interface), none alone are comprehensive.
- In part, this is because much of the advanced functionality of R comes from hundreds of user contributed packages. Hunting for what you want can be time consuming, and it can be hard to get a clear overview of what procedures are available.

### Steep Learning Curve



### Another warning

- I am older than dirt. Well, not really but close.
- I used R before it existed and was Splus.
- So since I am a dinosaur, I use R in a very rudimentary fashion.
- Our hip, young TFs use R all the time for their research and have more modern skills-so go to section to learn more about R.

### R's Ups and Downs

#### Plusses

- ☐ Completely free, just download from Internet
- □ Runs on many operating systems
- ☐ Many add-on packages for specialized uses
- □ Open source

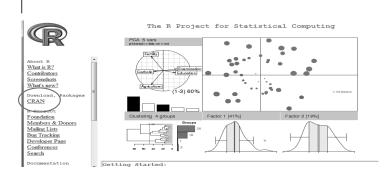
#### Minuses

☐ Obscure terms, intimidating manuals, odd symbols, inelegant output (except graphics)

### | Installing R

- www.r-project.org/
- download from CRAN
- select a download site
- download the base package at a minimum
- download contributed packages as needed
- Then (maybe) download RStudio: http://www.rstudio.com/

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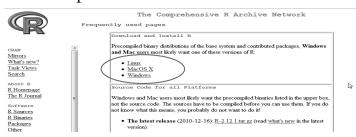
### Select a Mirror

#### ■ This is were you download from. USA!USA!

http://cran.cnr.Berkeley.edu http://cran.stat.ucla.edu/ http://cran stat ucla edw/ bttp://streaming.stat.isastate.edw/CRAN/ bttp://ftp.uss.gi.n.edw/CRAN/ bttp://rwatson.nci.nih.gov/cran mirror/ bttp://cran natu.edw/ bttp://cran case.edw/ bttp://cran case.edw/ bttp://cran case.edw/ bttp://cran case.edw/ bttp://cran case.edw/ bttp://bb.stat.crun.edw/R/CRAN/ bttp://stat.orun.edw/R/CRAN/ bttp://stat.orun.edw/R/CRAN/ bttp://stat.orun.edw/R/CRAN/ http://cran.mirrors.hoobly.com http://mirrors.nics.utk.edu/cran http://cran.revolutionanalytics.com http://cran.fhcrc.org/

University of California, Berkeley, CA University of California, Los Angeles, CA Iowa State University, Ames, IA Iowa State Cineversity, Ames, IA
Indiana University of Kansas, Lawrence, KS
National Cancer Institute, Bethesda, MD
Michigan Technological University, Houghton, MI
Washington University, St. Louis, MO
Case Western Reserve University, Cleveland, OH
Oregon State University
Staffib, Carnegie Mellon University, Pittsburgh, PA
Hooby Classifieds, Pittsburgh, PA
Motional Institute for Computational Sciences, Oak Ridg
Revolution Analytics, Dallas, TX
Fred Hitchinson Cancer Research Center, Seattle, WA
Western Washinaton University. Bellineham. WA Indiana University

# Precompiled Binaries



### **Tutorials**

- From R website under "Documentation"
  - ☐ "Manual" is the listing of official R documentation
    - An Introduction to R
    - R Language Definition
    - Writing R Extensions
    - R Data Import/Export
    - R Installation and Administration
    - The R Reference Index
- The Class website has several good introductory tutorials which is really all you need.

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# | Learning R



### Actually Running R

■ What will happen when you click on the R icon?

R x64 3.2.5

3.2.5 is the version number....you might download a newer version.

might download a newer version.

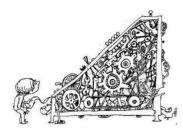
### The Basic R Screen



## The R prompt (>)

- > This is the "R prompt."
  - ☐ It says R is ready to take your command.
- R is an <u>interpretative language</u>, anything you type in it tries to interpret for you.

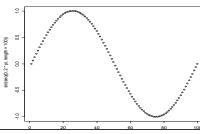
### R is an overgrown calculator



```
> ((2+6/3)*5^2)/4
[1] 25
> pi
[1] 3.141593
> log(pi)
[1] 1.14473
> sin(.5)/cos(.3)*sqrt(10)
[1] 1.586956
```

### R is a graphing calculator

> plot(sin(seq(0, 2\*pi, length=100)))



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### R can do vectors

- A vector is a sequence of values, all of the same type
- c() function returns a vector containing all its arguments in order

```
> students = c("Sean", "Louisa", "Frank", "Farhad", "Li")
> midterm = c(80, 90, 93, 82, 95)
```

■ Typing the variable name at the prompt causes it to display

```
> students
[1] "Sean" "Louisa" "Frank" "Farhad" "Li"
```

### Indexing

■ vec[1] is the first element, vec[4] is the 4th element of vec

```
> students
[1] "Sean" "Louisa" "Frank" "Farhad" "Li"
> students[4]
[1] "Farhad"
```

■ vec[-4] is a vector containing all but the fourth element

```
> students[-4]
[1] "Sean" "Louisa" "Frank" "Li"
```

-

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### Vector Arithmetic

Operators apply to vectors "pairwise" or "elementwise":

```
> midterm = c(80, 90, 93, 82, 95) #midterm exam scores
> final = c(78, 84, 95, 82, 91) # Final exam scores
> (midterm+final)/2
[1] 79 87 94 82 93
> course.grades = 0.4*midterm + 0.6*final
> course.grades
[1] 78.8 86.4 94.2 82.0 92.6
```

### Another Example Vectorized Math

```
> weight = c(110, 180, 240) ## create vector of weights
> height = c(5.5, 6.1, 6.2) ## create vector of heights
> bmi = (weight*4.88)/height^2 ## divides element wise
> bmi
[1] 17.74545 23.60656 30.46826
> > x=1:10
> x
    [1] 1 2 3 4 5 6 7 8 9 10
> sum(x)
[1] 55
> prod(x)
[1] 3628800
```

### Pairwise Comparisons

■ Is the final score higher than the midterm score?

> final>midterm
[1] FALSE FALSE TRUE FALSE FALSE

■ Boolean operators can be applied elementwise

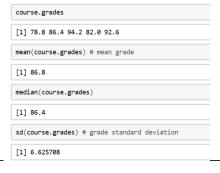
> (final < midterm) & (midterm > 80)
[1] FALSE TRUE FALSE FALSE TRUE

### Functions on Vectors

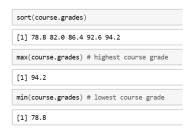
Command	Description
sum(vec)	sums up all the elements of vec
mean(vec)	mean of vec
median(vec)	median of vec
min(vec), max(vec)	the largest or smallest element of vec
sd(vec), var(vec)	the standard deviation and variance of vec
length(vec)	the number of elements in vec
pmax(vec1, vec2), pmin(vec1, vec2)	example: pmax(quiz1, quiz2) returns the higher of quiz1 and quiz2 for each student
sort(vec)	returns the vec in sorted order
order(vec)	returns the index that sorts the vector vec
unique(vec)	lists the unique elements of vec
summary(vec)	gives a five-number summary
any(vec), all(vec)	useful on Boolean vectors

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### | Functions on Vectors



### More Functions on Vectors



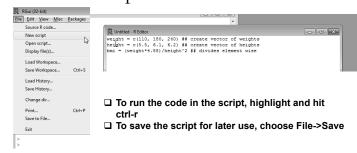
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### STOP: Understand Workflow

Work-flow
/'werkflo/
noun
the sequence of industrial, administrative, or other processes through which a piece of work passes
from initiation to completion.

- Its important to works smart with R and keep your commands in a file for easy access [for editing, sending to people, using again and again].
- Rstudio is even fancier with this-see video on class web site.

### Use an R Script File



### Phew...Using R: Creating a Data Set

- > Scores = c(22, 34, 18, 29, 36)
  - c means "concatenate" in R
  - in plain English "treat as data set"
- Now do:
  - >Scores

R will print the data set

### Important Rules

- 1. We created a variable
- 2. Variable names are case sensitive
- 3. No blanks in name

(can use \_ or . to join words, but not -)

4. Start with a letter (capitol or lowercase)

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### Non-numeric Data

- Enclose in quotes, single or double
- Separate entries with comma
- Example:

> names = c("Mary", "Tom", "Ed", "Dan", "Meg")

### Finding objects you created

To list the objects that you have in your current R session use the function ls().

> ls()

[1] "x" "y"

If you assign a value to an object that already exists then the contents of the object will be overwritten with the new value (without a warning!). Use the function rm to remove one or more objects from your session.

> rm(x, y)

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# Saving Stuff

- To exit: quit() or q() [the function quit!]
- Brings up this screen:



- Do what you want: Yes or No
  - Do Yes,
  - □ then re-open R, get Scores & names

### Using R Functions: Simple Stuff

Commands for mean, sd, summary (NB: function names case sensitive)

- mean(Scores)
- □ sd(Scores)
- summary(Scores)
- We will review some basic statistics next class.

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### R functions

- A zillion of 'em
- R's big strength, most common use
- For examples:
  - □Help
    - Google "+R +cran +standard deviation"
    - Enter name of a function [e.g., help(sd)]
  - ☐ Yields lots (!) of information

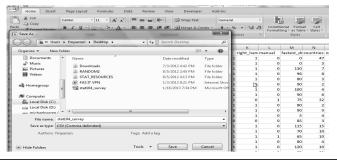
### Importing Data

- How do we get data into R?
- Remember we have no point and click...
- First make sure your data is in an easy to read format such as CSV (Comma Separated Values).
- In Excel, you can use the "save-as" command to convert any file to csv format.

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### CSV from Excel



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### First we need to get the filename

- fname = file.choose()
- The file.choose command opens a window in



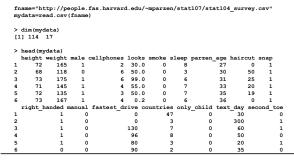
### Now we need to read it in

#### ■ mydata = read.csv(fname)

> fname=file.choose() > mydata=read.csv(fname) > dim(mydata) [1] 114 17 > names (mydata) [1] "height" [5] "looks" "weight" "male" "cellphones" "smoke" "sleep" "parzen\_age"
"manual" [9] "haircut" "snap" "fastest drive" "countries' "only\_child" "text day' "second\_toe"

This data is from a Stat 104 class survey.....

### Read data in from the internet



#### R commands are vectorized

- Vectorized sounds like a fancy word.
- It just means you can do a lot of things at once:

> mean (mydata)				
height	weight	male	cellphones	looks
67.85087719	147.75438596	0.51754386	4.06140351	30.27807018
smoke	sleep	parzen age	haircut	snap
0.05263158	6.74561404	45.24561404	34.67543860	0.89473684
right_handed	manual	fastest_drive	countries	only_child
0.90350877	0.26315789	90.14912281	8.40350877	0.1666667
text_day	second_toe			
72.88157895	0.36842105			
> sd(mydata)				
height	weight	male	cellphones	looks
4.9977561	27.5107578	0.5018983	1.9968342	24.6447889
smoke	sleep	parzen_age	haircut	snap
0.2242827	1.2305938	6.6339985	29.1257559	0.3082471
right_handed	manual	fastest_drive	countries	only_child
0.2965673	0.4422915	28.9002656	7.4592928	0.3743234
text_day				
282.2299736	0.4845061			

### Working with data.

- Accessing columns.
- The variable **mydata** has our data in it.... But you can't see it directly.
- Well you can, but it's a huge 114x17 matrix

> d	im (mydat	ca)								
[1]	114 17	7								
> m	ydata									
	height	weight	male	cellphones	looks	smoke	sleep	parzen_age	haircut	snap
1	72	165	1	2	30.0	0	8.0	27	0	1
2	68	118	0	6	50.0	0	3.0	30	50	1
3	73	175	1	6	99.0	0	6.0	31	25	1
4	71	145	1	4	55.0	0	7.0	33	20	1
5	72	135	1	3	50.0	0	7.0	35	19	1
6	73	167	1	4	0.2	0	6.0	36	0	1
7	72	155	1	5	25.0	0	6.0	37	3	1

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# Selecting Columns

■ To select a column use mydata\$column.

> names (mydata) [1] "height" "weight"																											
																	"cel										
[5]						"sı	noke	e"				sle					"par			ge"							
[9] "haircut" ":			"sr	nap'	"			"right handed"						"mar	ual	L"											
[13]	"fas	stes	t c	lriv	re"	"cc	ount	trie	es"		"0	only	r ch	nilo	1"		'tex	t	lay"								
[17]	"sec	cond	l Ec	e"									_					_									
> myd	ata	hei	ght																								
[1]	72	68	73	71	72	73	72	67	71	69	71	75	63	72	79	63	65	54	67	70	65	54	69	66	64		
[26]	63	69	66	68	72	74	63	64	66	72	68	69	67	73	71	69	68	67	69	64	80	71	71	75	53		
[51]	71	75	64	68	72	68	73	66	68	74	69	68	68	70	64	64	70	65	63	70	73	66	70	76	63		
[76]	66	61	67	63	74	75	63	70	74	69	69	63	73	70	69	62	62	72	67	59	50	67	70	68	70		
[101]	63	60	64	64	72	64	68	66	62	72	68	71	70	63													
> myd	ata	sle	ep																								
[1]	8.0	3.	0 6	5.0	7.0	7.	.0	6.0	6.0	0 6	.0 1	3.0	7.0	0 7.	.0 7	7.0	7.0	8 (	.0 6	5.0	7.0	5.	. 0	4.0			
[19]	7.0	5.	0 5	6.0	6.0	5.	.0 !	5.0	7.0	8 0	.0	7.0	8.0	0 7	.0 7	7.0	7.0	7	. 0 .	7.0	7.0	7	.0	6.0			
[37]	7.5	5 7.	0 7	1.0	7.0	8.	. 0 .	7.0	8.5	5 7	.0	в. о	8.0	0 7.	.0 7	7.0	8.0	6	.0 8	3.0	7.0	8 (	.0	7.5			
[55]	7.0	7.	0 7	1.0	7.0	7.	.0 (	6.0	7.0	7	.0 !	5.0	7.0	0 6	.0 8	3.0	8.0	8 (	.0 7	7.0	7.0	8 (	.0	6.0			
[73]	7.0	7.	0 7	1.0	7.0	8.	.0 8	в.о	7.0	7	.0	6.5	5.0	8.0	.0 5	5.5	5.0	6	.0 7	7.0	7.0	7.	.0	в.о			
[91]	7.0	4.	0 5	5.0	6.0	9.	.0 8	в.о	7.0	9	. 0	6.0	7.0	9.	.0 5	5.0	8.0	7	.0 6	6.5	8.0	7.	. 0	5.0			
[109]	6.0	6.	0 7	1.0	4.0	7.	. 0 :	1.0																			

### Selecting Columns

- One can also explicitly ask for different columns of the data matrix
- mydata[,5] returns all rows of the fifth column
- mydata[1,] returns the first row, all columns
- mydata[1:10,] returns the first ten rows
- mydata[,5:7] returns all rows for columns 5-7.
- head(mydata) first few rows of data
- tail(mydata) last few rows of data

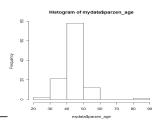
# Working with data.

- Subsetting data.
- Use a logical operator to do this.
  - $\square$  ==, >, <, <=, >=, <> are all logical operators.
  - □ Note that the "equals" logical operator is two = signs.
- Example
  - $\square$  mydata[mydata\$male == 1,] (note the comma!)
  - $\ \square$  This will return the rows where Gender is male (coded 1).
  - □ Remember R is case sensitive!
  - $\hfill \square$  This code does nothing to the original dataset.
  - $\hfill \square$  mydata.male = mydata[mydata\$male == 1,] gives a dataset with the appropriate rows.

## Basic Graphics

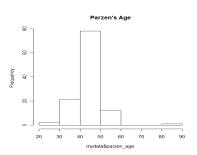
#### ■ Histogram

 $\square$  hist(mydata\$parzen\_age)



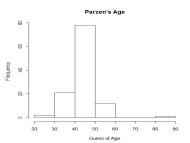
### Basic Graphics

- Add a title...
  - The "main" statement will give the plot an overall heading.
  - ☐ hist(mydata\$parzen\_age ,main="Parzen's Age")
  - ☐ Pro tip-use arrow keys to repeat previous commands, then edit the previous command.



### Basic Graphics

- Adding axis labels...
- Use "xlab" and "ylab" to label the X and Y axes, respectively.
- hist(mydata\$parzen\_age,
  main="Parzen's
  Age",xlab="Guess of
  Age", ylab="Frequency")

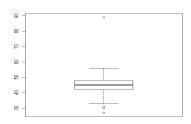


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### **Basic Plots**

#### ■ Box Plots

boxplot(mydata\$parzen\_age)



### Boxplots

#### ■ Change it!

boxplot(mydata\$parzen\_age
,main="Boxplot of
Parzen's Age",ylab="Guess
of Age")

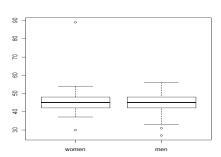


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# Box-Plots - Groupings

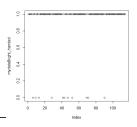
- What if we want several box plots side by side to be able to compare them
- First Subset the Data into separate variables.
  - ☐ mydata.m=mydata[mydata\$male==1,]
  - ☐ mydata.w=mydata[mydata\$male==0,]
- Then Create the box plot.
  - ☐ boxplot(mydata.m\$parzen\_age,mydata.f\$parzen\_age)
  - □ boxplot(mydata.w\$parzen\_age,mydata.m\$parzen\_age, names=c("women","men"))

# Boxplots – Groupings



### Discrete Data Plots

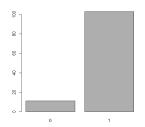
- We have a variable for whether someone is right handed
- > plot(mydata\$right\_handed)



Discrete Data Plots

> counts=table(mydata\$right\_handed)
> counts

0 1 11 103 > barplot(counts)

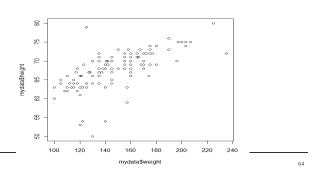


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### Scatter Plots

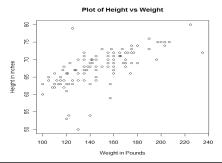
- Suppose we have two variables and we wish to see the relationship between them.
- A scatter plot works very well.
- R code:
  - $\square$  plot(x,y)
- Example
  - $\square$  plot(mydata\$weight,mydata\$height)

Scatterplots



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### Scatterplots



plot(mydata\$weight,mydata\$height,xlab="Weight in Pounds;;, vlab="Height in inches", main="Plot of Height vs Weight")

### Line Plots

- Often data comes through time.
- We will cover the quantmod package more in a little bit.
- Consider \$AAPL stock
  - > library(quantmod)
- [1] "AAPL"
- > head(AAPL)

	AAPL.Open	AAPL.High	AAPL.Low	AAPL.Close	AAPL.Volume	AAPL.Adjusted	
2007-01-03	86.29	86.58	81.90	83.80	309579900	11.19449	
2007-01-04	84.05	85.95	83.82	85.66	211815100	11.44295	
2007-01-05	85.77	86.20	84.40	85.05	208685400	11.36147	
2007-01-08	85.96	86.53	85.28	85.47	199276700	11.41757	
2007-01-09	86.45	92.98	85.15	92.57	837324600	12.36603	

### Line Plots



### R Packages

- One of the strengths of R is that the system can easily be extended.
- The system allows you to write new functions and package those functions in a so called `R package' (or `R library'). The R package may also contain other R objects, for example data sets or documentation.
- Just a few examples, there are packages for portfolio optimization, drawing maps, exporting objects to html, time series analysis, spatial statistics and the list goes on and on.

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# The QUANTMOD Package

- From the package description:
  - The quantmod package for R is designed to assist the quantitative trader in the development, testing, and deployment of statistically based trading models.

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### The Library Command

- Remember, once the package has been installed, it isn't automatically loaded into R.
- To use the package, you need to specify it via the library command. This has to be done once per session.
- Syntax: library(quantmod) [or whatever the package name is]

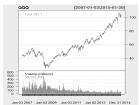
### An Introductory Package

- There is a new R package called Swirl that runs an introduction to R inside R.
- See http://swirlstats.com/students.html for instructions
- install.packages("swirl")
- library(swirl)
- swirl()

### Quantmod Example (more next time)

> library(quantmod)
> getSymbols("QQQ")
[1] "QQQ"

chartSeries(QQQ, theme=chartTheme('white')



Quantmod Example

library(quantmod)
getSymbols("QQQ",from="2015-01-01")
qqq.a=adjustOHLC(QQQ)

chartSeries(qqq.a, theme=chartTheme('white'))



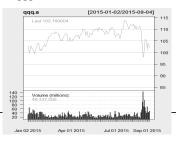
This is bar plot showing the range of each day



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### Quantmod Graph

- One can change the type of chart using the "type=" switch. Typical R behavior
- > chartSeries(qqq.a, theme=chartTheme('white'),type="1")



### What does getSymbols do?

■ Returns a matrix with the desired data

```
> dim(qqq.a)
[1] 171 6
> head(qqq.a)
```

QQ.Open QQQ.High QQQ.Low QQQ.Close QQQ.Volume QQQ.Adjusted 2015-01-02 103.5119 103.9609 102.20493 102.70379 31148800 102.70378 2015-01-05 102.2648 102.3745 100.90791 101.19725 36521300 101.19725 2015-01-06 101.3469 101.5165 99.39141 99.84037 66205500 99.84037 2015-01-07 100.5088 101.3669 100.25940 101.12741 37577400 101.12741 2015-01-08 102.0154 103.2625 101.87569 103.06296 40212600 103.06296 2015-01-09 103.3722 103.4122 101.78589 102.38452 41410100 102.38452

getSymbols is an unusual function since it automatically (and silently) creates an object.

### Let's review this object

■ There are some shortcut commands

Op(), Vo(), Ad(), Cl(),Hi(),Lo()

```
> tail(Op(qqq.a))
           QQQ.Open
2015-08-28
             105.08
2015-08-31
             105.03
2015-09-01
             101.68
2015-09-02
2015-09-03
             104.32
2015-09-04
             101.97
> tail(Cl(qqq.a))
           QQQ.Close
2015-08-28
              105.62
2015-08-31
2015-09-01
              101 05
2015-09-02
              103.90
2015-09-03
              103 39
2015-09-04
              102.16
```

### Paneling Graphics

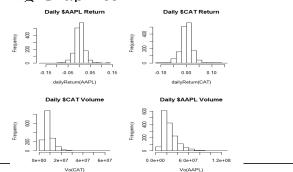
- Suppose we want more than one graphic on a panel.
- We can partition the graphics panel to give us a framework in which to panel our plots.

### Paneling Graphics

#### Consider the following

```
> par(mfrow=c(2,2))
> hist(dailyReturn(AAPL),main="Daily $AAPL Return")
> hist(dailyReturn(CAT),main="Daily $CAT Return")
> hist(Vo(CAT),main="Daily $CAT Volume")
> hist(Vo(AAPL),main="Daily $AAPL Volume")
```

### Paneling Graphics



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# Functions

- We have seen a lot of built-in functions: mean(), sd(), plot(), read.csv()...
- An important part of programming and data analysis is to write custom functions
- Functions help make code modular
- Functions make debugging easier
- Remember: this entire class is about applying functions to data

### A Really Simple Function

■ A really simple function

```
addOne = function(x) {
  x+1
}
```

■ This function adds 1 to its input

> addOne(10)
[1] 11

0

#### \_\_\_\_\_

### Another function example

#### Three number summary function

```
threeNumberSummary = function(x) {
    c(mean=mean(x), median=median(x), sd=sd(x))
}
> x <- rnorm(100, mean=5, sd=2) # Vector of 100 normals with mean 5 and sd 2
> threeNumberSummary(x)
    mean median sd
4.766603 4.754881 1.944234
```

### R Function Example

#### ■ Code

```
mysquare = function(x)
{
  cat("The square of ",x,"is",x*x,"\n")
}
```

#### ■ What do you think this function does?

\_ \_\_\_\_

# Running the Function

```
> mysquare(5)
The square of 5 is 25
> mysquare(10)
The square of 10 is 100
> mysquare("mike")
Error in x * x : non-numeric argument to binary operator
> mysquare(mike)
The square of 3 is 9
> mike
[1] 3
```

### | Modifying the Function

■ This function now also returns a value

```
mysquare = function(x)
{
  cat("The square of ",x,"is",x*x,"\n")
  return(x*x)
}
```

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# Running the new function

```
> mysquare(5)
The square of 5 is 25
[1] 25
> mike
[1] 3
> mike=mysquare(5)
The square of 5 is 25
> mike
[1] 25
```



### Things you should know

☐ How to install R

☐ How to read csv files into R

☐ How to load packages into R (quantmod)

☐ How to write very simple functions in R