

DATA SOCIETY®

“If you can’t explain it simply, you don’t understand it well enough.”

- Albert Einstein

Course syllabus

1. What is data science?
2. Manipulating data in R
3. Visualization in R



Setting expectations

Data science takes dedication! You will need to:

1. Take this course 😊
2. Practice
3. Review class material on your own
4. Practice
5. Complete exercises outside of class
6. Practice
7. Share and read latest news

Outline

- What is data science?
- A data scientist's approach
- Introduction to R
 - Calculations in R
 - Reading data into R
 - Manipulating data in R
- Visualization in R
 - Basic plotting

How is data being used?

Retail	Finance	Marketing	Real estate	Cool
Target:	Kabbage:	Netflix:	Zillow:	Andrew Ng:
The store knows you're pregnant based on what you buy	Makes lending decisions based on Amazon product reviews, etc.	What movie should you watch?	Calculates Zestimate (value of your home)	Machine learning techniques recognize cat faces online using pictures and videos
				

What is “Big Data”?

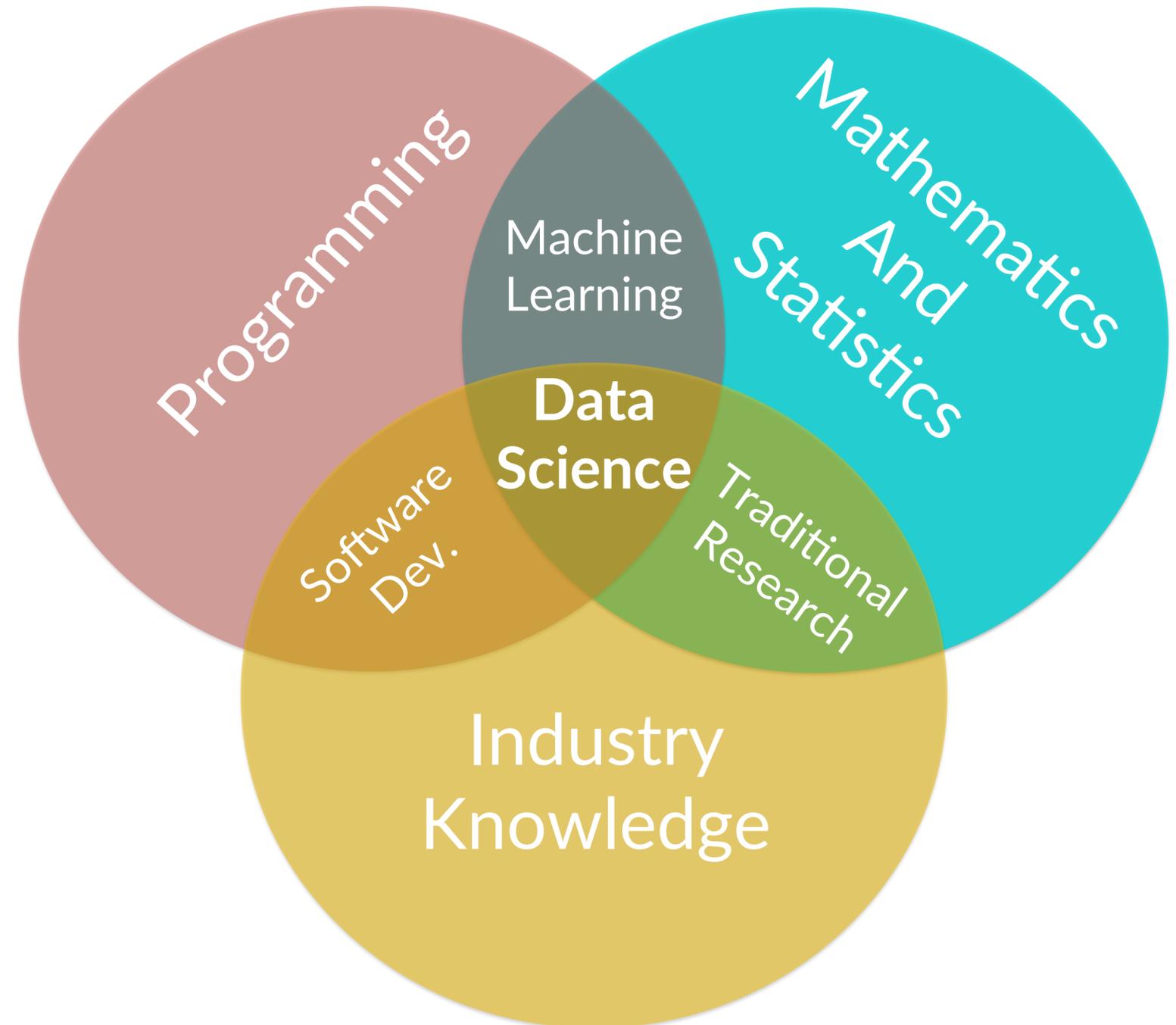
- Big Data is large volumes of information
 - Moving
 - Storing
 - Manipulating
 - Accessing
- It is not:
 - Analysis or insights



That's why you're in this class!

What is data science?

- Data science applies the scientific method to analyzing data
- It lies at the intersection of several disciplines
- It draws on domain specific knowledge that makes the analysis of Big Data possible



Who is a data scientist?

- An analyst who can:
 1. **Pose** the right question
 2. **Wrangle** the data (gather, clean, and sample data to get a suitable data set)
 3. **Manage** the data for easy access by the organization
 4. **Explore** the data to generate a hypothesis
 5. **Make predictions** using statistical methods such as regression and classification
 6. **Communicate** the results using visualizations, presentations, and products

Levels of expertise

Data analyst

- **Wrangles** the data
- **Manages** the data
- **Creates** basic analyses and visualizations

Data modeler

- **Models** to answer specific questions
- **Understands the data**, its source and structure

Data scientist

- **Asks** the right questions
- **Looks** for patterns in data
- **Interprets** results critically



Data science job market

Somewhat important ✓

Very important ✓

	A non-data-driven company	The business is just starting to collect data	Data is the product of the company	Company uses data to make decisions
Basic tools	✓	✓	✓	✓
Software engineering		✓	✓	✓
Statistics	✓	✓	✓	✓
Machine learning			✓	✓
Data processing		✓	✓	✓
Data visualization and communication	✓	✓	✓	✓
Thinking like a data scientist	✓	✓	✓	✓

Who hires data scientists?



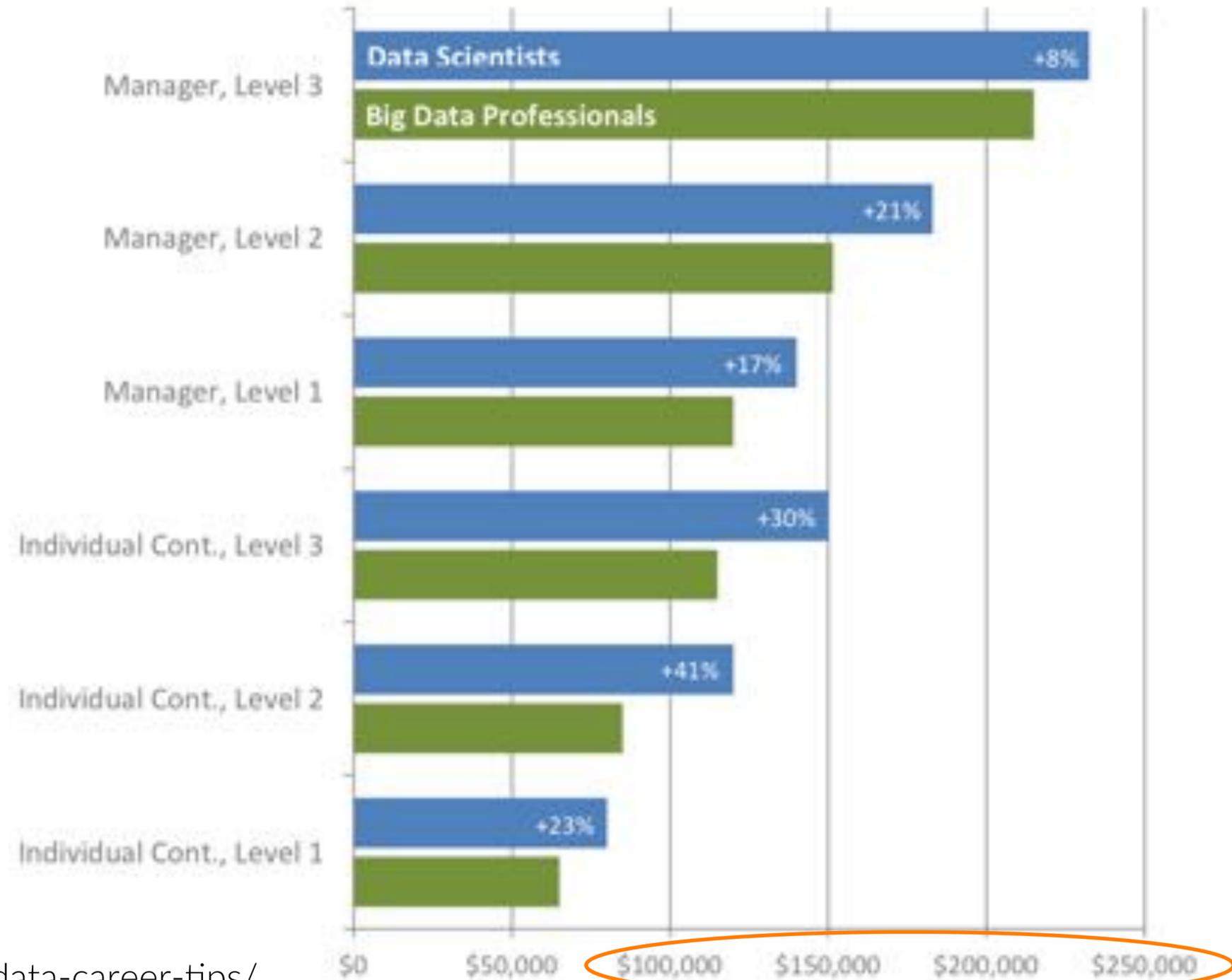
Source: datasciencecentral.com

How much do data scientists make?

- According to a Burtch Works 2014 data science job market survey:

“Data scientists earn a median salary that can be up to 40% higher than predictive analytics professionals at the same job level”

- The graphic on the right provides detail on median salaries by experience level



Source: <http://www.burtchworks.com/big-data-analyst-salary/big-data-career-tips/>

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Data science control cycle



Data science control cycle



For every job there is a tool

Data storage

- Hadoop
- Spark
- SQL
- ...



Data analysis

- R
- SAS
- SPSS
- Matlab
- Python
- Google Prediction API
- ...

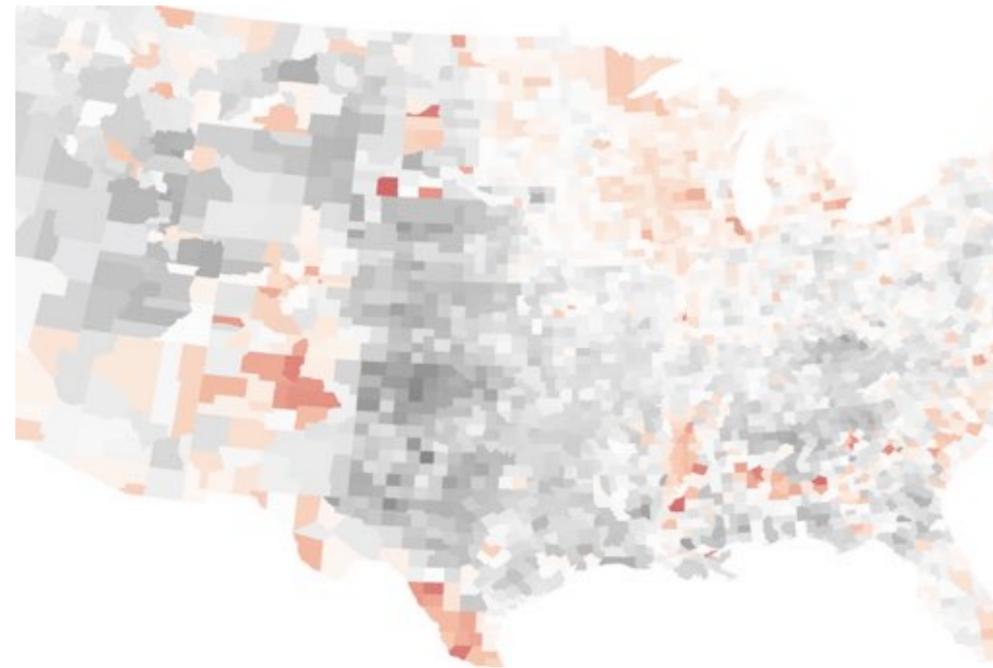
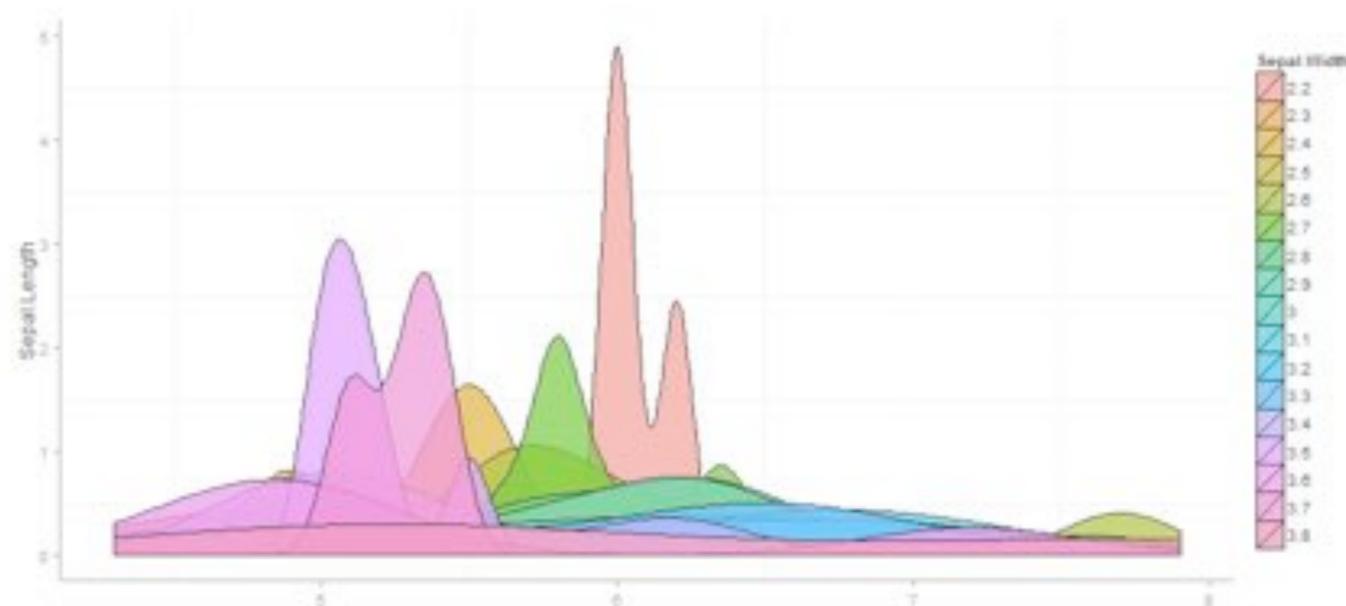


For every job there is a tool

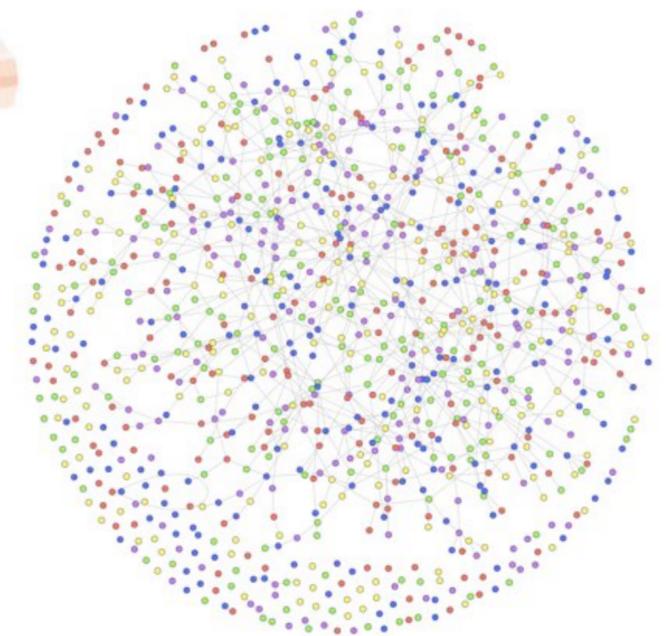
Visualization

- R
- Tableau
- iVEDiX

Density plot



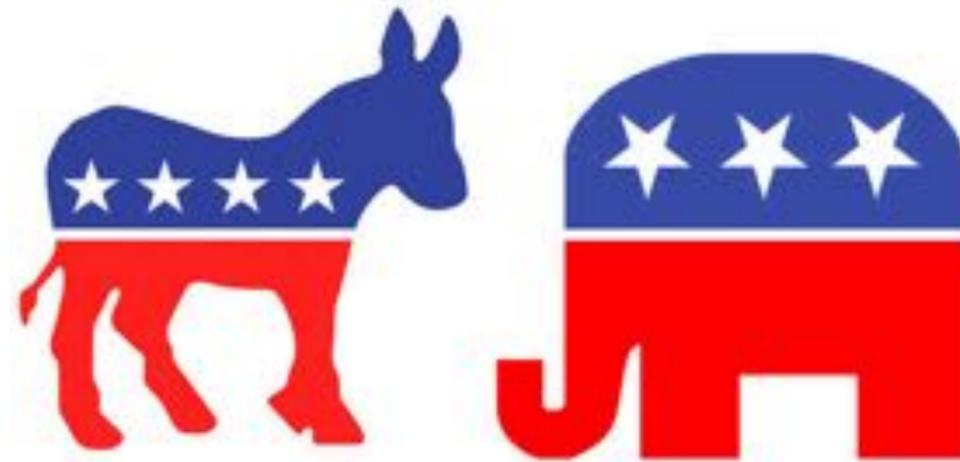
All of these visualizations were created in R



Supervised machine learning

Pattern discovery when inputs (x) and outputs (y) are known

Input x:
Voter



Output y:
Political
affiliation

Examples: Classification and regression are supervised machine learning

Unsupervised machine learning

The data inputs (x) have no target outputs (y)

Input x :
Voter



Output y :
Not given
(to be discovered)

We want to impose structure on the inputs (x) to say something meaningful about the data

Machine vs. human

	Machine	Human
Understanding context		✓
Thinking through the problem		✓
Asking the right questions		✓
Selecting the right tools		✓
Performing calculations quickly	✓	
Performing repetitive tasks	✓	
Following pre-defined rules	✓	
Interpreting results		✓

Setting up R: overview

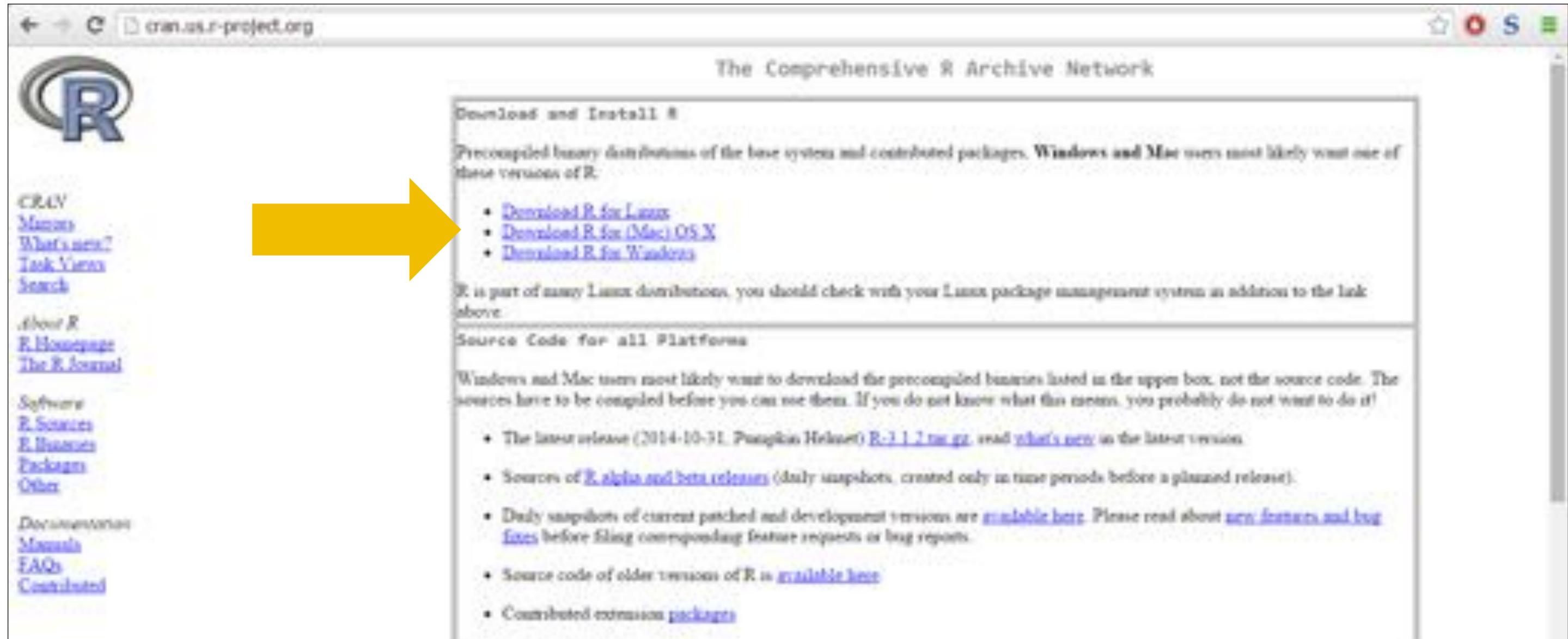
1. What is R?
2. Download R from the CRAN website (<http://cran.us.r-project.org/>)
 - R for Windows
 - R for Mac
3. Install R Studio (<http://www.rstudio.com/products/rstudio/download/>)
 - RStudio a brief tour
4. Running a script
 - Variables
5. Reading in a data
 - Manually
 - Through the script

What is R?

- R is a statistical programming software
 - Has many similar features to SAS, Excel and SPSS
 - Scripting language
- It's free and open source
- Has lots of helpful pre-built functions
 - You can build your models quicker
- Easy to learn



Install R



The screenshot shows the CRAN website (cran.us.r-project.org) with the title "The Comprehensive R Archive Network". The main content area is titled "Download and Install R" and contains the following text:

Precompiled binary distributions of the base system and contributed packages. **Windows and Mac users** most likely want one of these versions of R:

- [Download R for Linux](#)
- [Download R for \(Mac\) OS X](#)
- [Download R for Windows](#)

R is part of many Linux distributions, you should check with your Linux package management system in addition to the link above.

Source Code for all Platforms

Windows and Mac users most likely want to download the precompiled binaries listed in the upper box, not the source code. The sources have to be compiled before you can use them. If you do not know what this means, you probably do not want to do it!

- The latest release (2014-10-31, Puugkin Helmer) [R-3.1.2.tar.gz](#), read [what's new](#) in the latest version.
- Sources of [R alpha and beta releases](#) (daily snapshots, created only in time periods before a planned release).
- Daily snapshots of current patched and development versions are [available here](#). Please read about [new features and bug fixes](#) before filing corresponding feature requests or bug reports.
- Source code of older versions of R is [available here](#)
- Contributed extension [packages](#)

A yellow arrow points from the left side of the page towards the download links.

R for Windows



R for Mac

The screenshot shows the CRAN website for R on Mac OS X. The page title is "R for Mac OS X". The main content area contains the following text:

This directory contains binaries for a base distribution and packages to run on Mac OS X (release 10.6 and above). Mac OS 8.6 to 9.2 (and Mac OS X 10.1) are no longer supported but you can find the last supported release of R for these systems (which is R 1.7.1) [here](#). Releases for old Mac OS X systems (through Mac OS X 10.5) and PowerPC Macs can be found in the [old](#) directory.

Note: CRAN does not have Mac OS X systems and cannot check these binaries for viruses. Although we take precautions when assembling binaries, please use the normal precautions with downloaded executables.

R 3.1.2 "Pumpkin Helmet" released on 2014/10/31.

This binary distribution of R and the GUI supports 64-bit Intel based Macs on Mac OS X 10.6 (Snow Leopard) or higher.

Please check the MD5 checksum of the downloaded image to ensure that it has not been tampered with or corrupted during the mirroring process. For example type with `R-3.1.2-mavericks.pkg` in the Terminal application to print the MD5 checksum for the `R-3.1.2-mavericks.pkg` image. On Mac OS X 10.7 and later you can also validate the signature using `pkgutil --check-signature R-3.1.2-mavericks.pkg`

Files:

- [R 3.1.2 binary for Mac OS X 10.6 \(Snow Leopard\) and higher, signed package.](#) Contains R 3.1.2 framework, R.app GUI 1.65 in 64-bit for Intel Macs. The above file is an Installer package which can be installed by double-clicking. Depending on your browser, you may need to press the control key and click on this link to download the file.
- This package contains the R framework, 64-bit GUI (R.app) and Tcl/Tk 8.6.0 X11 libraries. The latter component is optional and can be omitted when choosing "custom install", it is only needed if you want to use the tcltk R package. GNU Fortran is **NOT** included (needed if you want to compile packages from sources that contain FORTRAN code) please see [the tools directory](#).
- [R 3.1.2 binary for Mac OS X 10.9 \(Mavericks\) and higher, signed package.](#) It contains the same software versions as above, but this R build has been built with Xcode 5 to leverage new compilers and functionalities in Mavericks not available in earlier OS X versions.

On the left side of the page, there is a navigation menu with the following items:

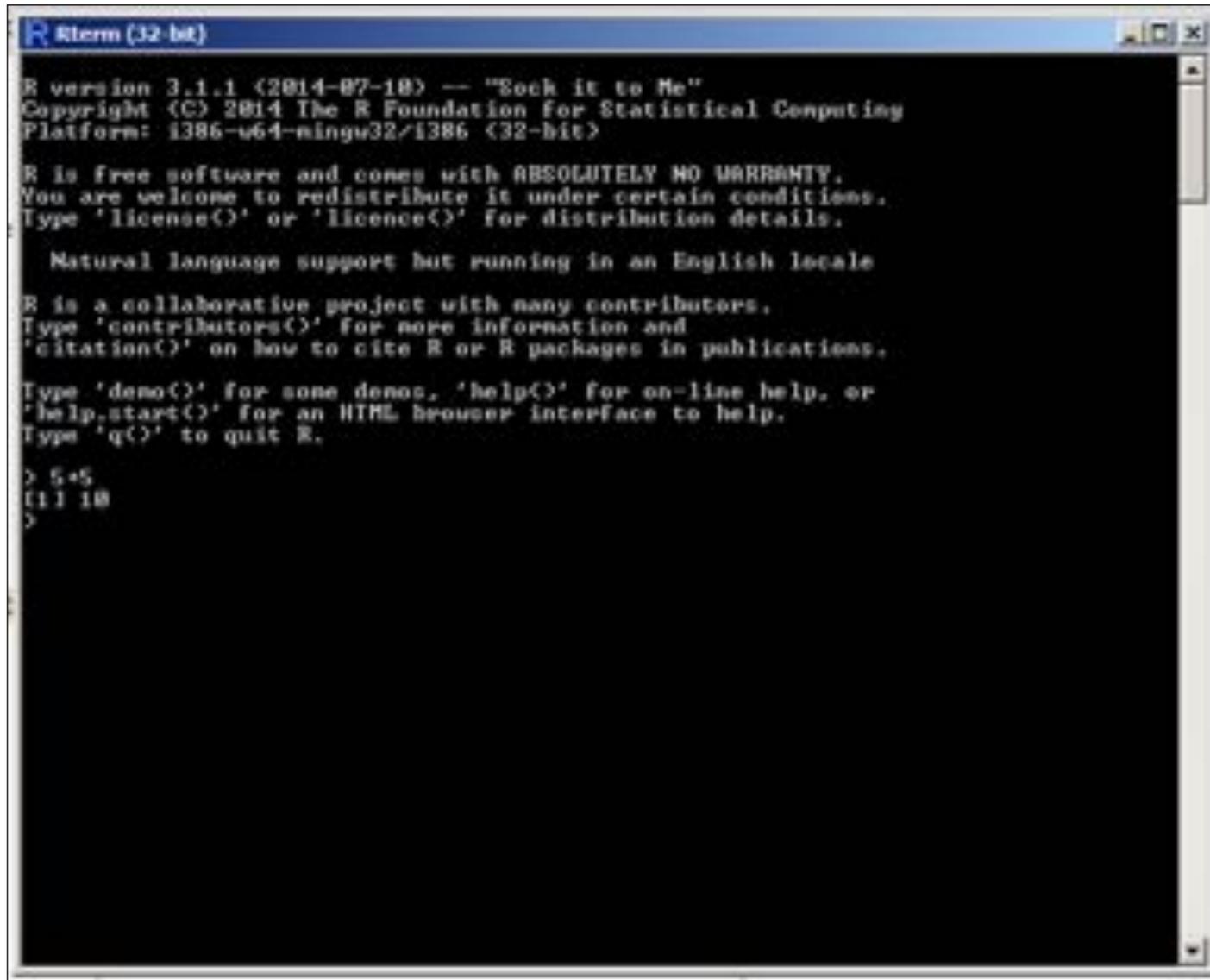
- CRAN
- Manual
- What's new?
- Task Views
- Search
- About R
- R Homepage
- The R Journal
- Software
- R Sources
- R Binaries
- Packages
- Other
- Documentation
- Manuals
- FAQs
- Contributed

A yellow arrow points from the "R Binaries" link in the navigation menu to the download link for R 3.1.2-snowleopard.pkg in the main content area.

What is RStudio?

The user interface in the R terminal is bulky and non-intuitive

RStudio provides a better interface for a more intuitive user experience



```
Rterm (32-bit)
R version 3.1.1 (2014-07-10) -- "Sock it to Me"
Copyright (C) 2014 The R Foundation for Statistical Computing
Platform: i386-u64-mingw32/i386 <32-bit>

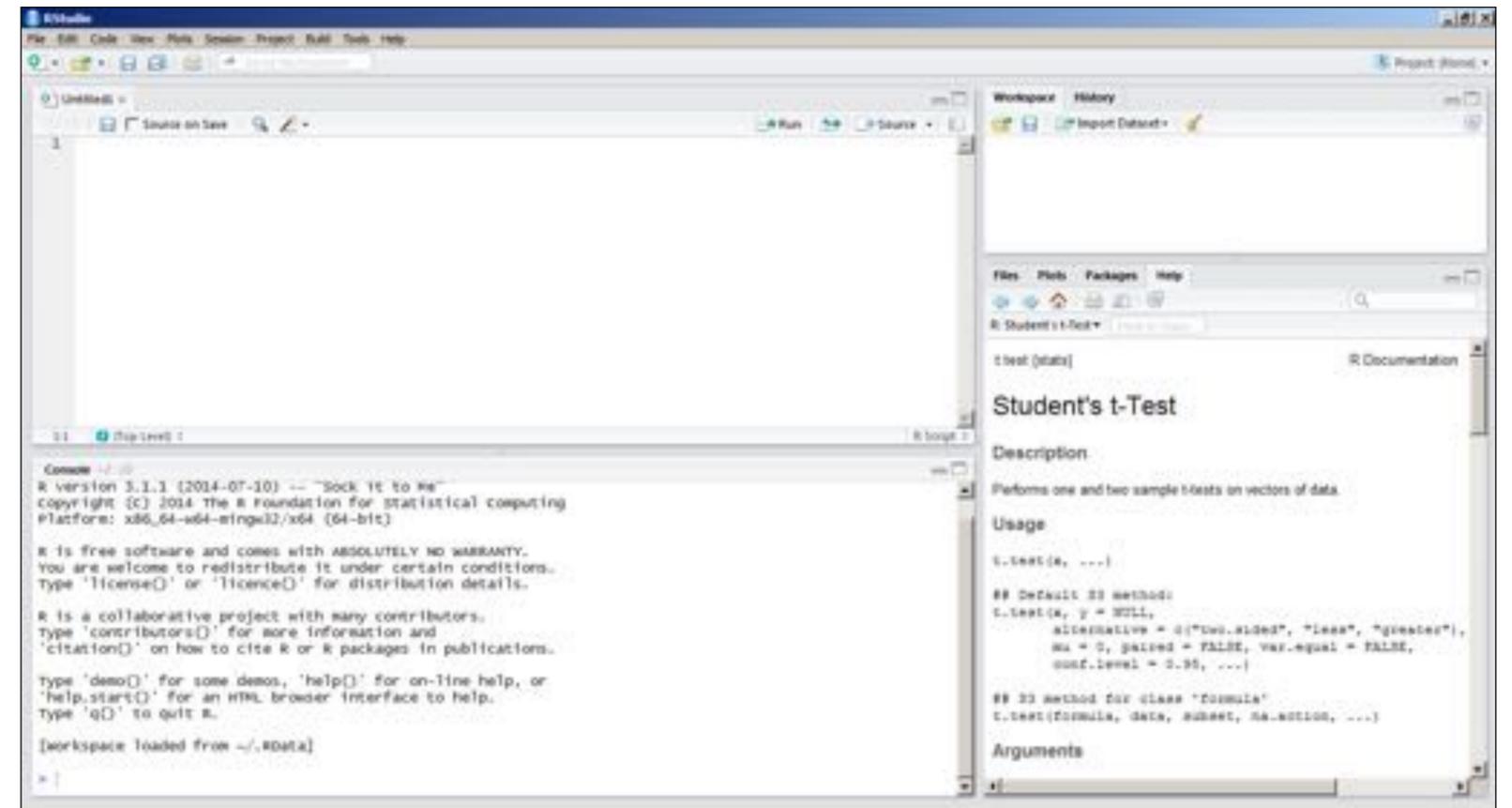
R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

Natural language support but running in an English locale

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> 5+5
[1] 10
>
```

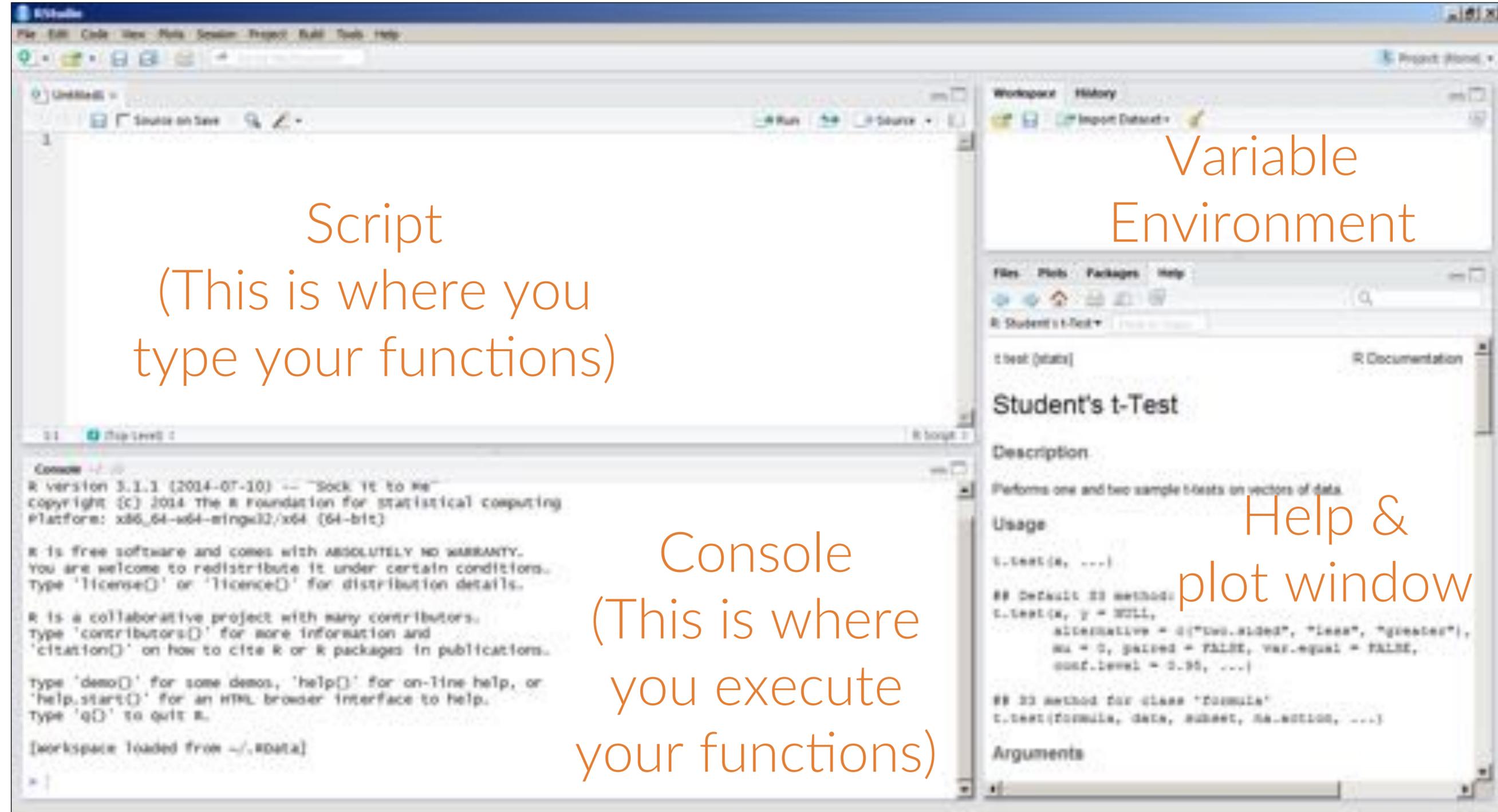


Install R Studio

The screenshot shows the RStudio website's download page. The URL in the browser is www.rstudio.com/products/rstudio/download/. The page features the RStudio logo and navigation links for Products, Resources, Pricing, About Us, and Blog. A prominent blue button on the right says "Click here to learn more about Shiny!". The main content area is titled "Download RStudio Desktop v0.98.1091 — Release Notes". Below this, there is a section for "Installers for ALL Platforms" which contains a table of download links for various operating systems. A large yellow arrow points to the first row of this table, which is for Windows.

Installer	Size	Date	MD5
RStudio v0.98.1091 - Windows (R/VS16/T3)	49 MB	2016-11-08	818f5a385c051151975da4085cad138128
RStudio v0.98.1091 - Mac OS X (10.9+)	36.4 MB	2016-11-08	9c7d2caa781c7478e4a776b79113483ee
RStudio v0.98.1091 - Debian 6+/Ubuntu 12.04+ (32-bit)	33 MB	2016-11-08	8bc579cbee43e534e37b4540955abeda
RStudio v0.98.1091 - Debian 6+/Ubuntu 12.04+ (64-bit)	34.9 MB	2016-11-08	1e88e4715993daa8c77d4d89f70e77e8
RStudio v0.98.1091 - Fedora 17+/RHEL 7+/openSUSE 11.4+ (32-bit)	33.4 MB	2016-11-08	5ee7812996188f98ec1c3712042788f
RStudio v0.98.1091 - Fedora 17+/RHEL 7+/openSUSE 11.4+ (64-bit)	33 MB	2016-11-08	6d2ac46ced731f5758f3d8e98125133

RStudio



Script
(This is where you
type your functions)

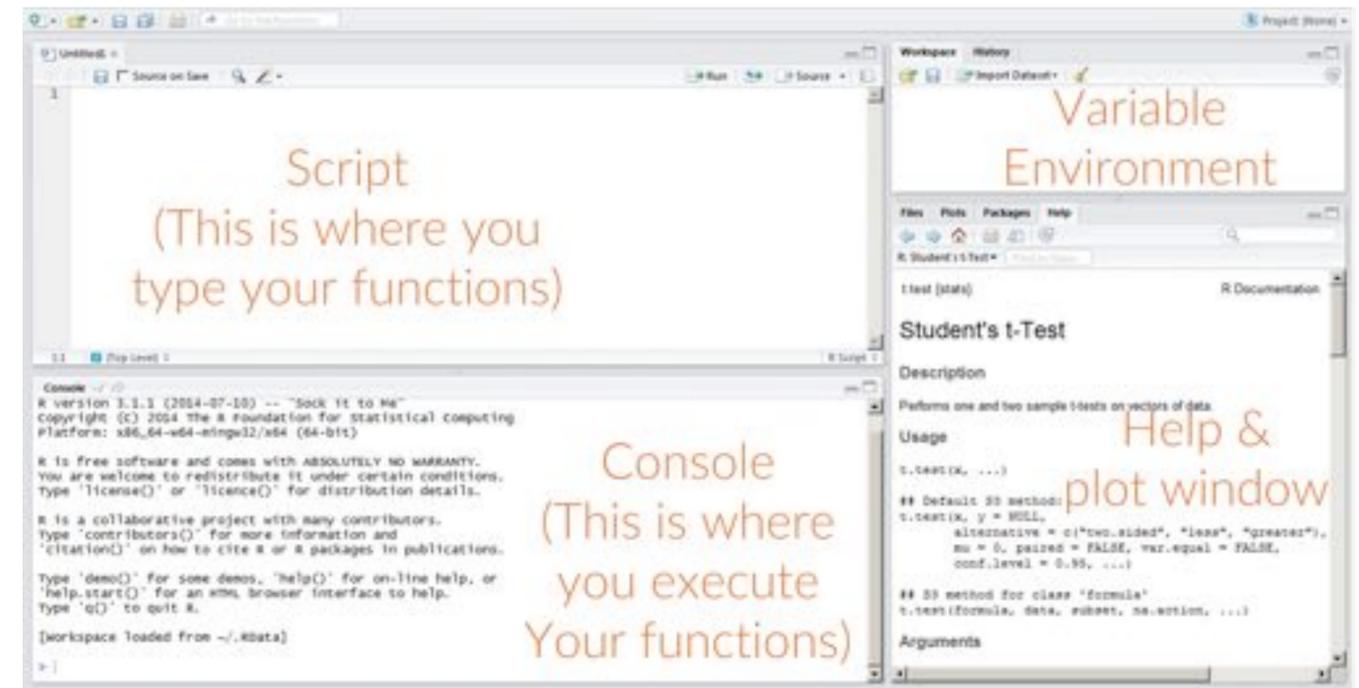
Variable
Environment

Console
(This is where
you execute
your functions)

Help &
plot window

Script and Console

- Script
 - We use scripts to generate re-usable code
 - They are like macros
 - Code is not executed here unless you press run (see next slide)
 - Good coding practice:
 - When saving a script use a name that describes what the script does i.e. counter
 - Comment your code using #
- Console
 - This is where the code executes
 - It is the actual R program (backend)
 - Anything that you type in here will be executed



Running code from script

The screenshot displays the RStudio environment. The main editor window, titled 'Untitled1.R', contains the following R code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6 |
7 #a=10
8 a+b*d
```

An orange rectangular box highlights the script editor area, with the text 'Type code in here' overlaid in orange. Below the script editor is the 'Console' window, which shows the R startup message and a prompt '>'. To the right, the 'Workspace' and 'History' panes are empty. Below them, the 'Files' pane shows the current project. The 'Help' pane is open to the 'Student's t-Test' documentation page, which includes a description and usage examples.

Running code from a script

The screenshot displays the RStudio interface. The main window shows a script editor with the following code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a=b+d
```

The code from line 3 to line 8 is highlighted in blue. An orange box surrounds the entire script editor window. Overlaid on the highlighted code is the text: "Highlight the code you want to run".

The console window at the bottom left shows the R startup message:

```
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

[workspace loaded from ~/.RData]
```

The right-hand pane shows the "Workspace" and "History" tabs, and the "Files" tab. The "Files" tab is active, showing the "Student's t-Test" documentation page.

Running code from a script

The screenshot displays the RStudio interface. The top-left pane shows an R script with the following code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a=b+d
```

An orange arrow points to the 'Run' button in the script editor's toolbar. Overlaid text reads: "To run the code in the console Press: Run or Hit: ctrl/command + enter".

The bottom-left pane shows the R console with the following output:

```
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

[workspace loaded from ~/RData]
```

The bottom-right pane shows the help window for the 'Student's t-Test' function, including a description: "Performs one and two sample t-tests on vectors of data." and usage information: "t.test(x, ...)" and "## Default 33 method: t.test(x, y = NULL, ...)".

Running code from a script

The screenshot displays the RStudio interface. The main editor window shows an R script with the following code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
```

The console window at the bottom shows the execution of this code:

```
> # Setting up variables
> # Remember things following # aren't executed
> a=3
> b=2
> d=4
>
> #a=10
> a+b+d
[1] 9
```

The Environment pane on the right shows the current workspace with the following variables:

Variable	Value
a	3
b	2
d	4

The Help pane on the right shows the documentation for the `t.test` function, titled "Student's t-Test".

Code is executed in the console

Running code from a script

The screenshot shows the RStudio interface. The main editor window contains the following R script:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
```

The console window at the bottom shows the execution of the script:

```
> # Setting up variables
> # Remember things following # aren't executed
> a=3
> b=2
> d=4
>
> #a=10
> a+b+d
[1] 9
```

An orange arrow points from the text "Notice #a=10 is a comment and not 'executed', otherwise a+b+d = 16 rather than 9" to the line `> #a=10` in the console. The right-hand pane shows the "Workspace" and "History" tabs, with the "Values" table displaying:

Variable	Value
a	3
b	2
d	4

The bottom-right pane shows the help page for "Student's t-Test", including a description: "Performs one and two sample t-tests on vectors of data." and usage information: `t.test(x, ...)`.

Variables

The screenshot displays the RStudio interface. The main editor window contains the following R code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
```

An orange arrow points from the code to the **Workspace** pane on the right, which shows the following table of values:

Variable	Value
a	3
b	2
d	4

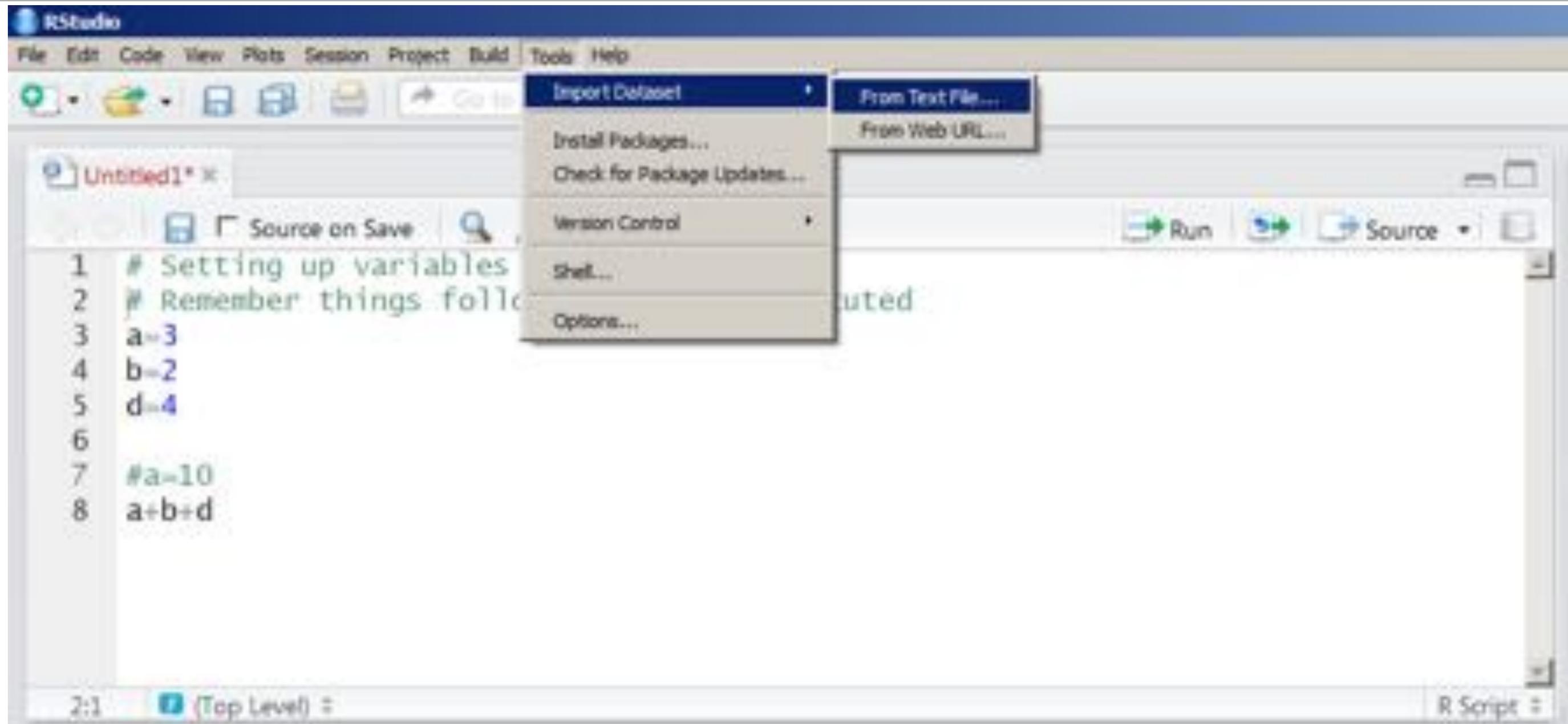
The **Console** pane at the bottom shows the execution of the code, with the final result of the expression `a+b+d` being `[1] 9`.

Below the code, the text reads: "Once the code is run the variables are 'stored' in memory. Here we have a,b,d".

Reading in data

- Data comes in many forms
 - R can read in most forms
- The most common to work with now days are
 - `csv` files – comma separated value files
 - `tsv` files – tab separated value files
- R can also read in others (excel and json) but that is not covered here

Manually



Tools > Import Dataset > From Text File...

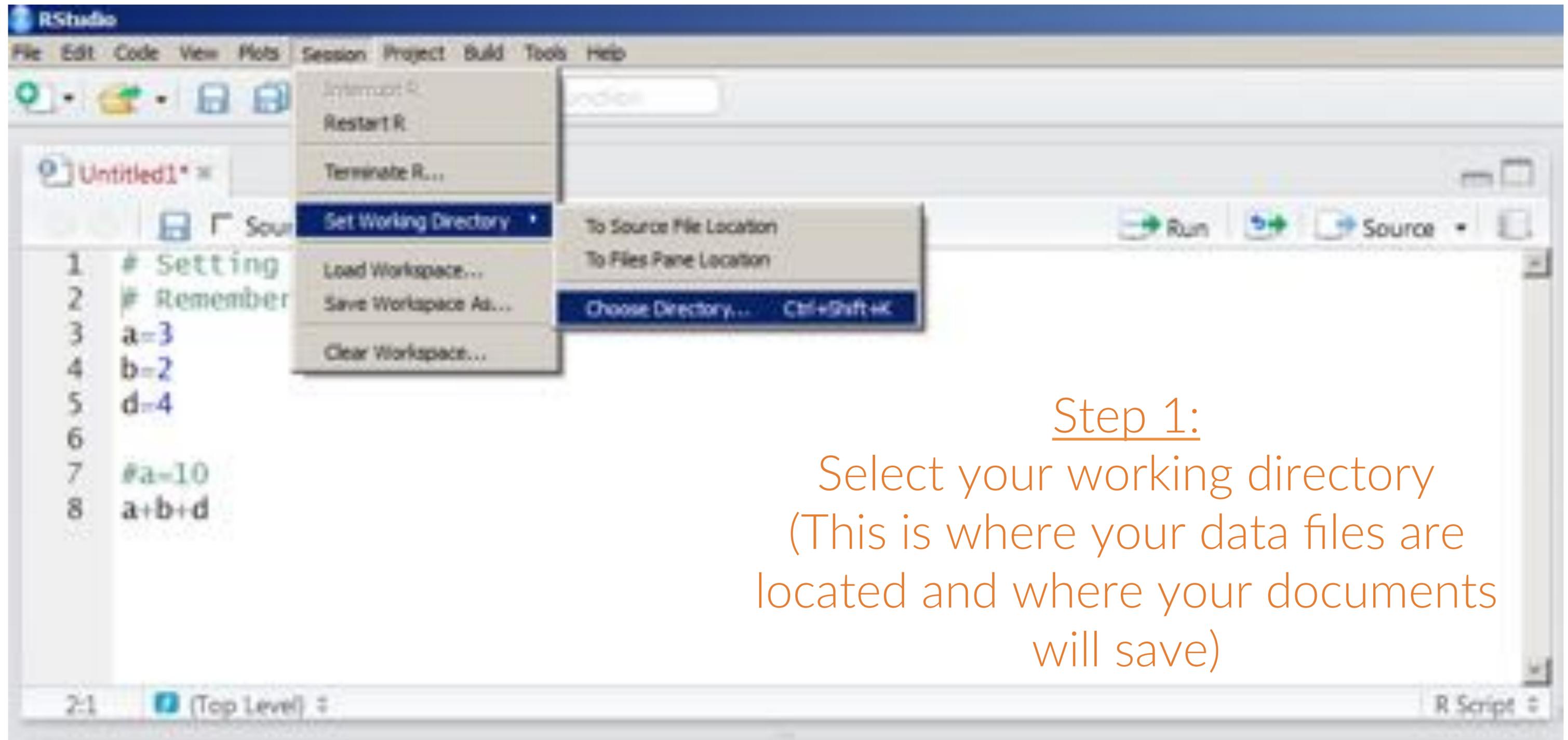
Manually

- R will automatically select options for you such as heading, sometimes it makes mistakes so be sure to double check

The screenshot shows the RStudio interface with the 'Import Dataset' dialog box open. The dialog is configured for a file named 'white_wine'. The 'Input File' field contains a long list of column names and values, including 'fixed.acidity', 'volatile.acidity', 'citric.acid', and 'residual.sugar'. The 'Data Frame' preview at the bottom shows a table with columns: fixed.acidity, volatile.acidity, citric.acid, and residual.sugar. The console on the left shows R code for setting up variables and a table of data.

fixed.acidity	volatile.acidity	citric.acid	residual.sugar
7.0	0.27	0.36	20.70
6.1	0.30	0.34	1.60
8.1	0.28	0.40	6.90
7.2	0.23	0.32	8.50
7.2	0.23	0.32	8.50
8.1	0.28	0.40	6.90
8.2	0.32	0.16	7.00
7.0	0.27	0.36	20.70
6.1	0.30	0.34	1.60
8.1	0.22	0.43	11.4
8.1	0.27	0.42	11.4
8.6	0.23	0.44	9.7
7.9	0.18	0.17	10.8
6.6	0.26	0.41	12.4
8.3	0.42	0.62	9.7
6.6	0.17	0.38	11.4
8.1	0.48	0.04	9.6

Through the Script

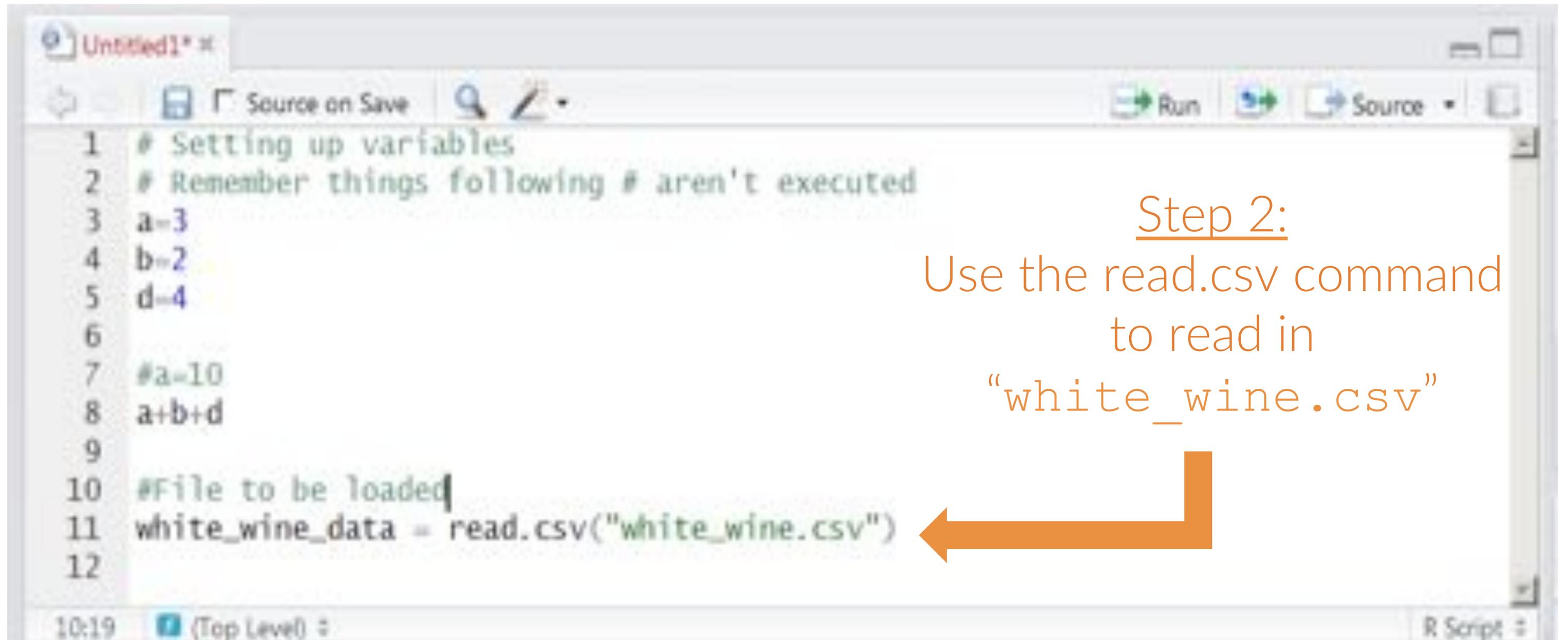


The screenshot shows the RStudio interface with the 'Session' menu open. The 'Set Working Directory' option is selected, and its sub-menu is visible, showing 'Choose Directory...' as the active option. The script editor on the left contains the following R code:

```
1 # Setting
2 # Remember
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
```

Step 1:
Select your working directory
(This is where your data files are located and where your documents will save)

Through the script



```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
9
10 #File to be loaded
11 white_wine_data = read.csv("white_wine.csv")
12
```

Step 2:
Use the read.csv command
to read in
"white_wine.csv"

Through the script



The screenshot shows an R script editor window titled "Untitled1*.R". The script contains the following code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b+d
9
10 #File to be loaded
11 white_wine_data = read.csv("white_wine.csv")
12
```

Annotations include:

- A curved orange arrow pointing from the "Run" button in the toolbar to the code on line 8.
- A text label "Step 3: Highlight and press run" in orange text, with a curved orange arrow pointing from the text to the "Run" button.
- A large orange arrow pointing from the right side of the editor to the code on line 11.

Through the script

The screenshot displays the RStudio interface. The main editor window shows an R script with the following code:

```
1 # Setting up variables
2 # Remember things following # aren't executed
3 a=3
4 b=2
5 d=4
6
7 #a=10
8 a+b=d
9
10 #File to be loaded
11 white_wine_data = read.csv("white_wine.csv")
12
```

The console window at the bottom left shows the execution of the script, with the command `white_wine_data = read.csv("white_wine.csv")` highlighted. An orange box around the console contains the text `<-Working directory location->`.

The workspace window on the right shows the object `white_wine_data` with 4898 observations and 12 variables. Below this, the values for variables `a`, `b`, and `d` are displayed:

Variable	Value
a	3
b	2
d	4

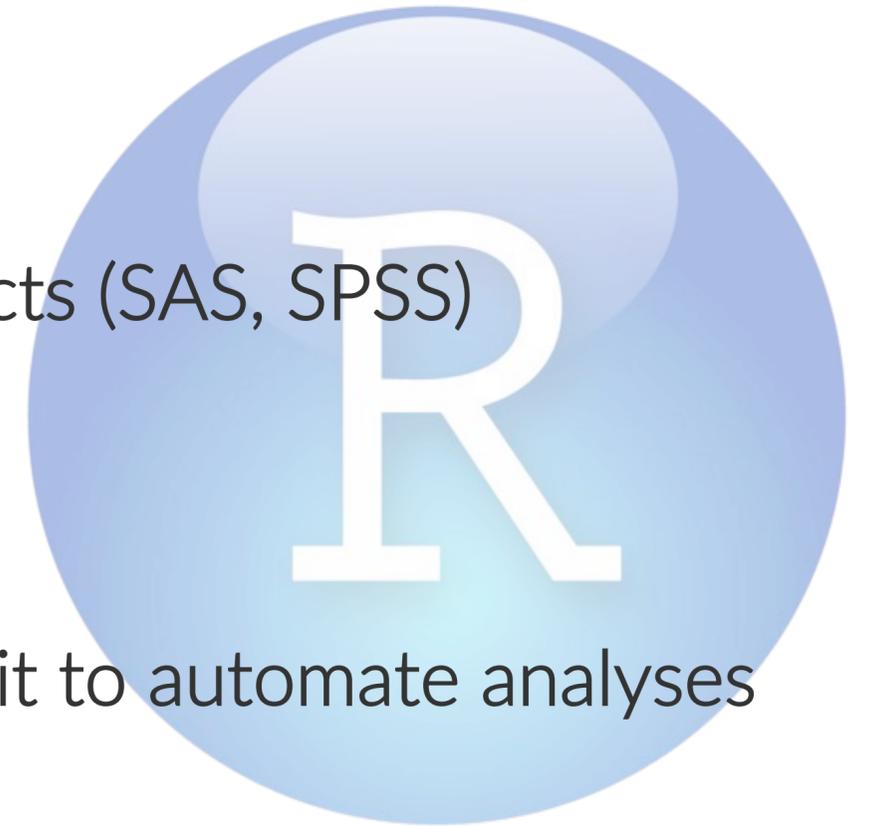
The bottom right pane shows the output of a Student's t-Test, with the title "Student's t-Test" and a "Description" section.

Outline

- What is data science?
- A data scientist's approach
- Introduction to R
 - Calculations in R
 - Reading data into R
 - Manipulating data in R
- Visualization in R
 - Basic plotting

Why use R

1. **De facto standard** among professional statisticians
2. Comparable and **often superior in power** to commercial products (SAS, SPSS)
3. Available for the Windows, Mac, and Linux operating systems
4. **R is a general-purpose programming language**, so you can use it to automate analyses
5. Create **dynamic graphics and visualization**
6. Large community of users, **many are prominent scientists**: www.r-bloggers.com
7. **Pre-made packages to run data analyses** contributed by user base (**over 6,500 packages**)



Source: <http://cran.r-project.org/web/packages/>

Uses of R

1. Can be used to analyze and visualize data
2. Can be used to write software
3. Can be used to create data products and applications

In this course, we will focus on how to analyze and visualize data



Companies that use R



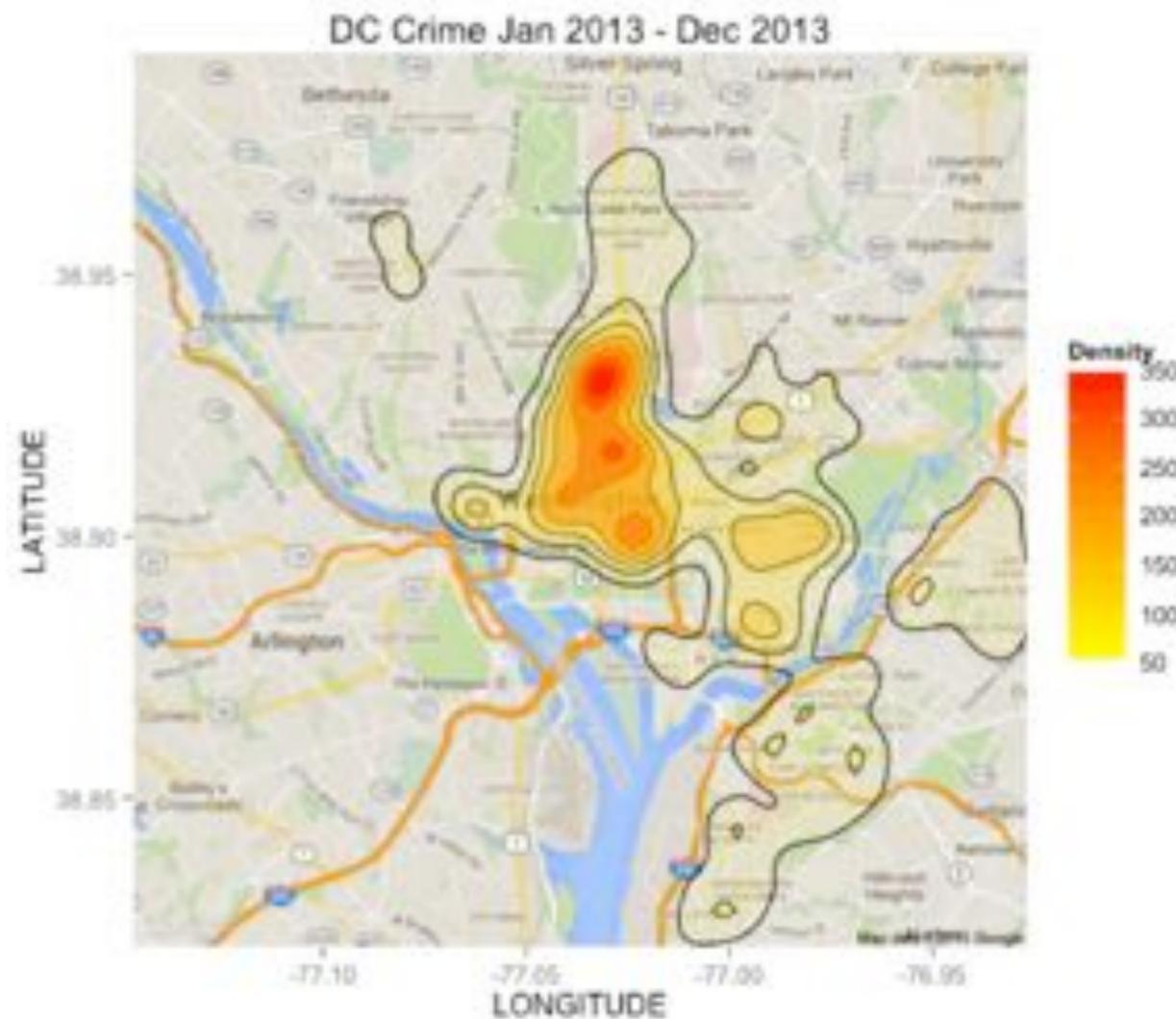
R vs. Excel

	R	Excel
Data capacity	R can read files as big as several gigabytes and trillions of data points; only limitation is your RAM	Excel can't read more than 1,048,576 rows and 16,384 columns (2011 version), files over ~300 megabytes can be very slow to work with
Customization	Can create custom visualizations through code, very flexible	Drop down menus limit ability to manipulate charts and graphs
Analyzing data	Powerful, pre-built packages that speed up work flow	Less flexible built-in analytic abilities that can be augmented by macros
Modeling	Data analysis and statistical models	Complex financial and accounting models
Seeing data	Built-in spreadsheet viewer	Easy to use spreadsheet interface
Usability	Direct commands similar to Excel "if-statements"	Keyboard shortcuts and slower point-and-click functionality

Visualizations in R

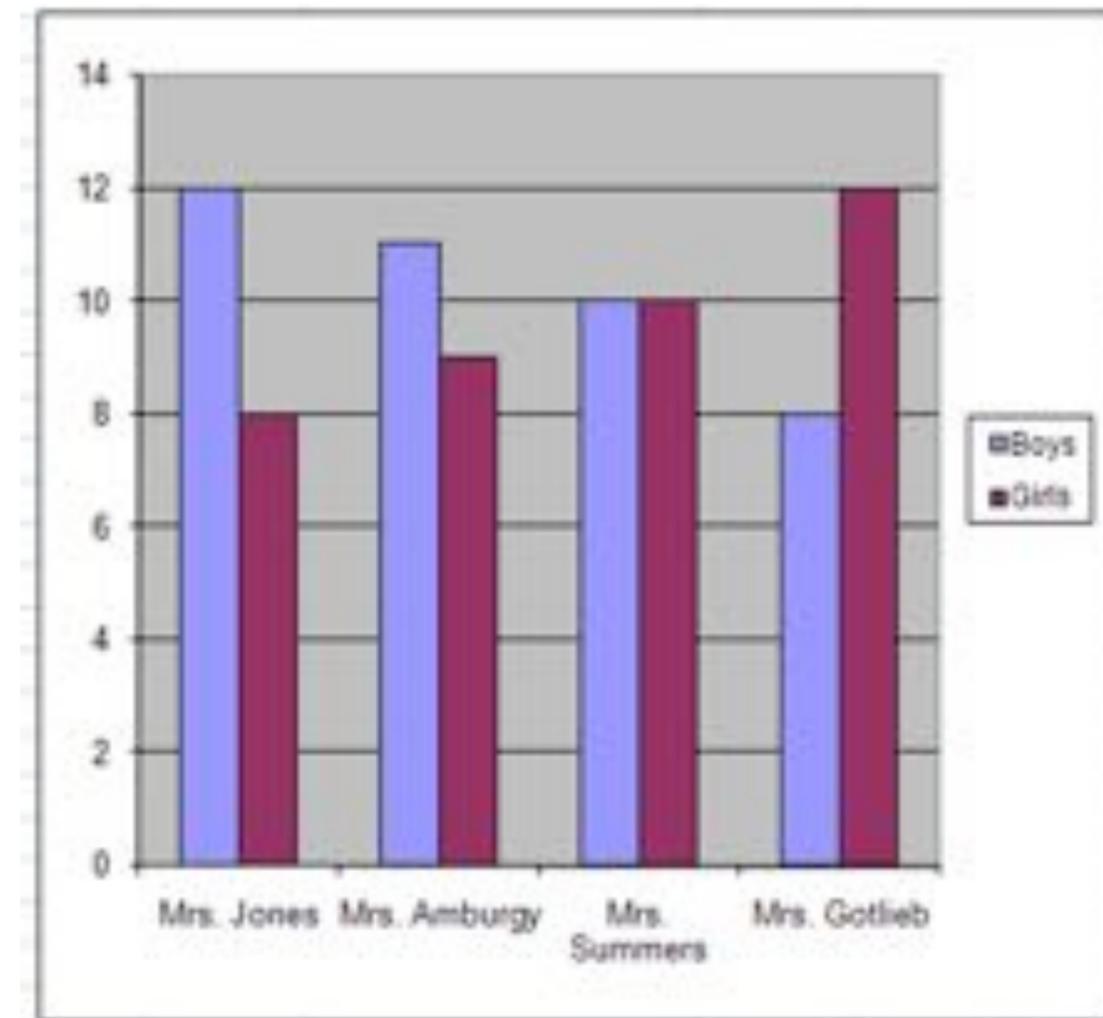
R

Simple customizable code: flexible



Excel

Drag and drop: rigid

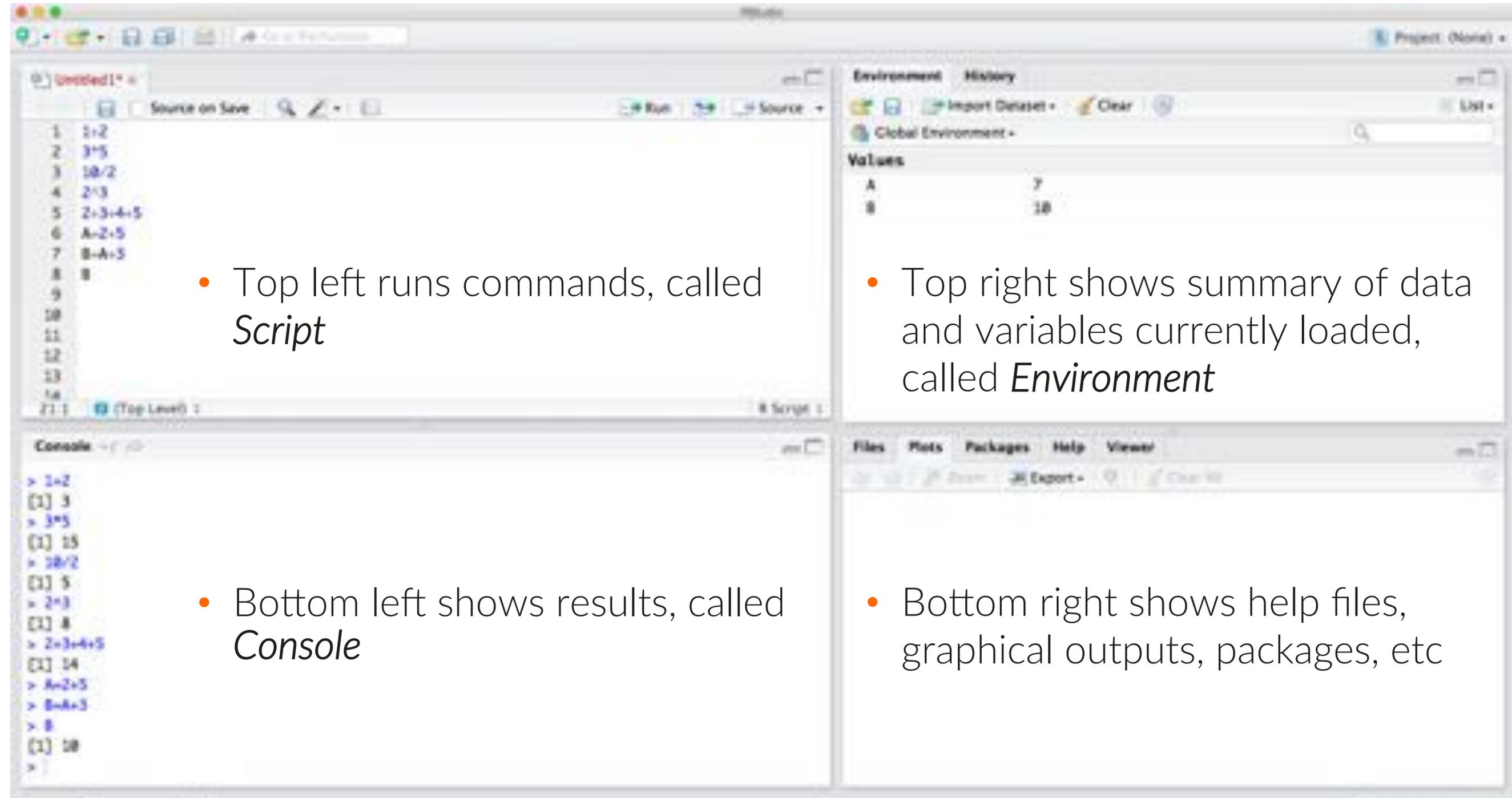


R vs. Python

- R has more convenient statistical packages to analyze data than Python
 - More than any other software tool, over 6,500 as of April 2015
- R is easier to learn for non-programmers than Python, less code is required to perform tasks
- Python is used by many data scientists to build data products (they also tend to be computer scientists)
- Python can be easier to integrate into web applications

Source: <http://cran.r-project.org/web/packages/>

RStudio overview



The screenshot displays the RStudio interface with four main panes:

- Top Left (Script):** Contains R code for arithmetic operations and variable assignments. The code is:

```
1 1+2
2 3*5
3 10/2
4 2*3
5 2+3+4+5
6 A=2+5
7 B=A+3
8 8
```
- Top Right (Environment):** Shows the current environment with variables A and B. The values are:

Variable	Value
A	7
B	10
- Bottom Left (Console):** Shows the output of the commands from the script. The output is:

```
> 1+2
[1] 3
> 3*5
[1] 15
> 10/2
[1] 5
> 2*3
[1] 6
> 2+3+4+5
[1] 14
> A=2+5
> B=A+3
> 8
[1] 10
>
```
- Bottom Right (Files/Plots/Packages/Help/Viewer):** Shows the help files, graphical outputs, packages, etc.

- Top left runs commands, called *Script*

- Top right shows summary of data and variables currently loaded, called *Environment*

- Bottom left shows results, called *Console*

- Bottom right shows help files, graphical outputs, packages, etc

Working with R: Comments

- Hashmarks are used to add comments and annotate your code

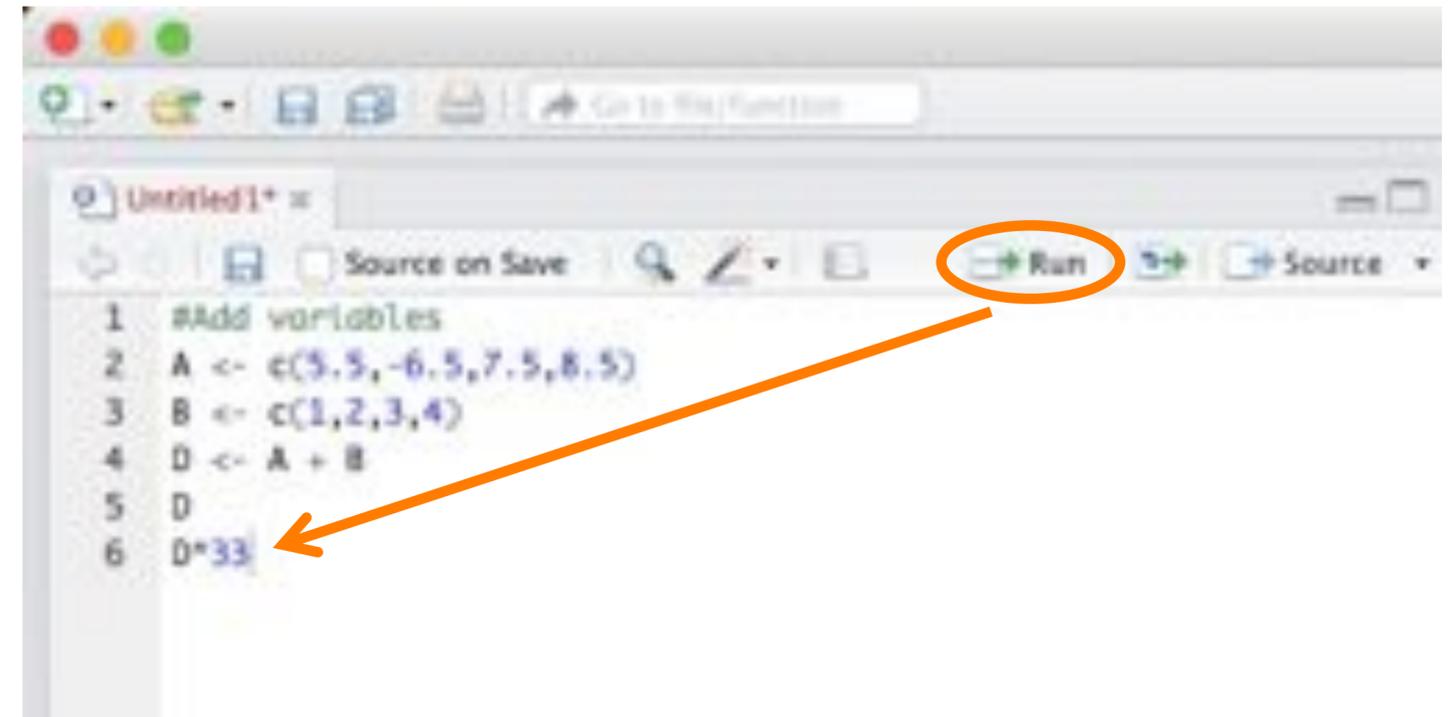
```
# Comments need to start with a hashmark, but don't need to end with one  
# Hashmarks show up in green and are included to explain your code
```

Script

- It's good practice to annotate your code
 - You can go back later and understand what you were doing

Executing commands in R

- Code is executed when you press “Run” in the top right hand corner of the script window
- R runs the line of code where your cursor is located
- You can also highlight multiple lines to run at once



Note: R is case sensitive

Outline

- What is data science?
- A data scientist's approach
- Introduction to R
 - Calculations in R
 - Reading data into R
 - Manipulating data in R
- Visualization in R
 - Basic plotting

Working with R: variables

- A series of numbers (think columns in Excel) can be defined using the arrow (\leftarrow) or equals ($=$) sign

```
# Define variables with arrow
A  $\leftarrow$  c(5.5, -6.5, 7.5, 8.5)
B  $\leftarrow$  c(1, 2, 3, 4)
```

Script

or

```
# Define variables with equals sign
A  $=$  c(5.5, -6.5, 7.5, 8.5)
B  $=$  c(1, 2, 3, 4)
```

Script

- The command `c ()` stands for “concatenate” (join) a series of numbers

Basic operations in R

Adding

- Just use + sign

```
Script
# Add variables
A = c(5.5, -6.5, 7.5, 8.5)
B = c(1, 2, 3, 4)
D = A + B
D
```

```
Console
> D
[1] 6.5 -4.5 10.5 12.5
```

Multiplying

- Just use * sign

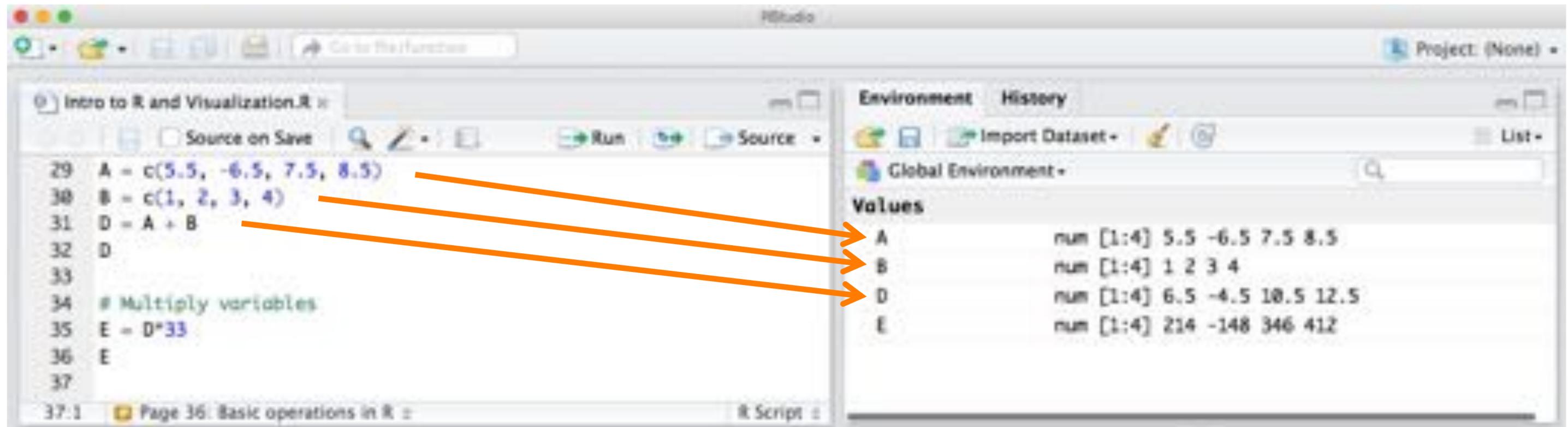
```
Script
# Multiply variables
E = D*33
E
```

```
Console
> E
[1] 214.5 -148.5 346.5 412.5
```

*Enter formulas in top left window (script)
Output is shown in bottom left window (console)*

Working with R: variables

- When a variable is named (instantiated), R stores it in its “environment” and can use it for subsequent operations



The screenshot shows the RStudio interface. The left pane displays the source code for a script named 'Intro to R and Visualization.R'. The code includes the following lines:

```
29 A = c(5.5, -6.5, 7.5, 8.5)
30 B = c(1, 2, 3, 4)
31 D = A + B
32 D
33
34 # Multiply variables
35 E = D*33
36 E
37
```

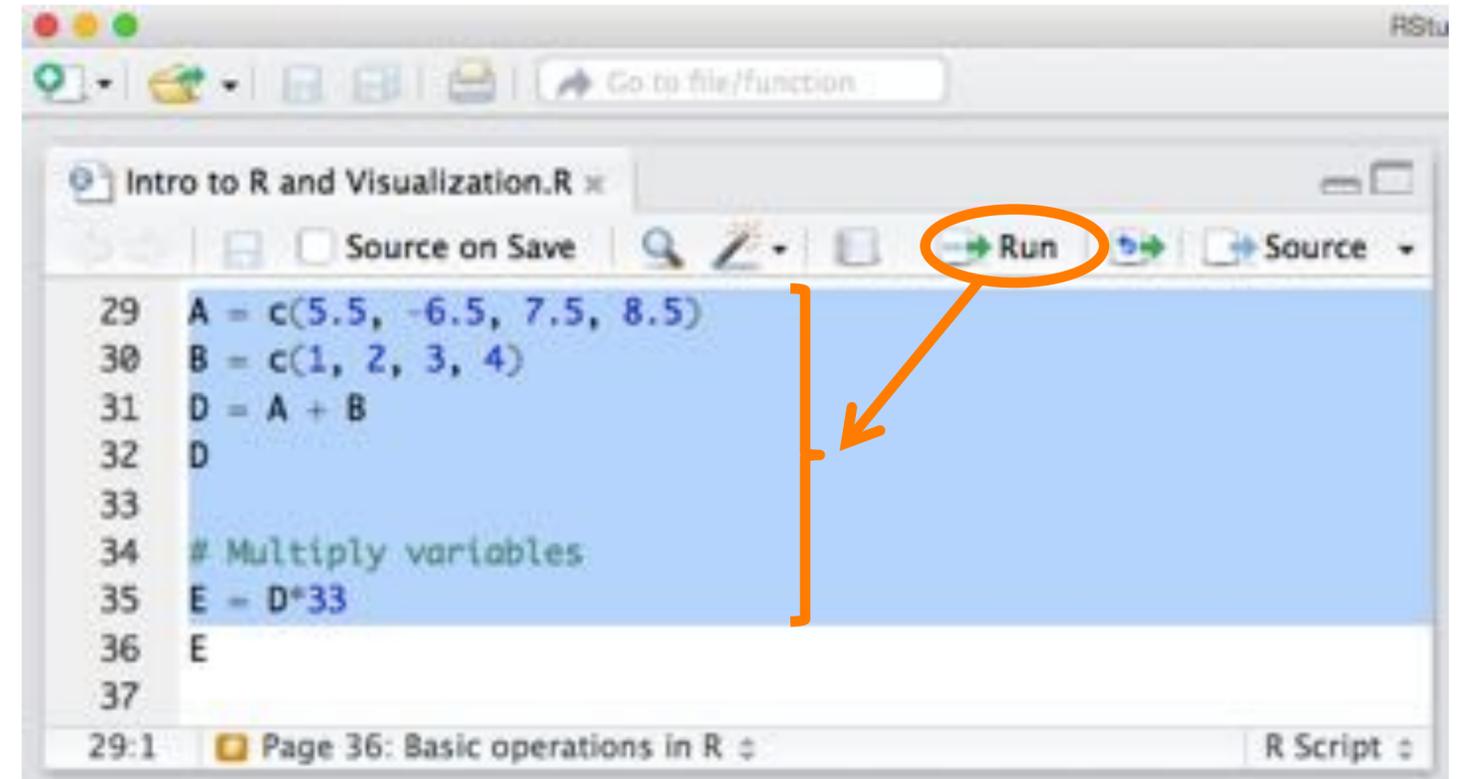
The right pane shows the Environment tab, which displays the values of the variables defined in the code:

Variable	Type	Value
A	num [1:4]	5.5 -6.5 7.5 8.5
B	num [1:4]	1 2 3 4
D	num [1:4]	6.5 -4.5 10.5 12.5
E	num [1:4]	214 -148 346 412

Three orange arrows point from the code lines to the environment table: one from line 29 to variable A, one from line 30 to variable B, and one from line 31 to variable D.

R can run several lines of code

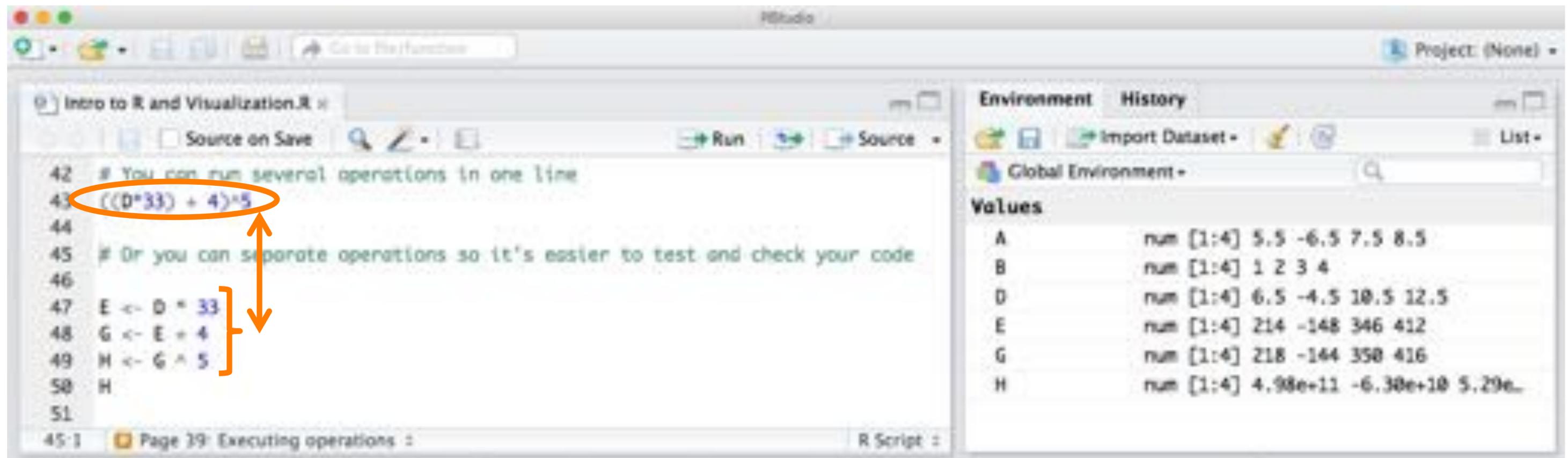
- You can highlight several lines of code and press “Run” to execute all of them
- Highlighting can be done either with the mouse or by holding “Shift” and using the arrow keys
- You can execute a command by pressing “Ctrl” + “Enter” for PCs or “Command” + “Enter” for Macs



Troubleshooting: if you have trouble with this, try restarting R, restarting your computer, or reinstalling R

Executing operations

- You can run several operations in 1 line of code
- Or you can separate steps and instantiate new variables to check your code more easily



The screenshot shows the RStudio interface. The source editor on the left contains the following R code:

```
42 # You can run several operations in one line
43  $((D*33) + 4)^5$ 
44
45 # Or you can separate operations so it's easier to test and check your code
46
47 E <- D * 33
48 G <- E + 4
49 H <- G ^ 5
50 H
51
```

An orange oval highlights line 43, and an orange arrow points from it to lines 47-49, which are grouped by a bracket. The Environment pane on the right shows the following values:

Variable	Type	Value
A	num [1:4]	5.5 -6.5 7.5 8.5
B	num [1:4]	1 2 3 4
D	num [1:4]	6.5 -4.5 10.5 12.5
E	num [1:4]	214 -148 346 412
G	num [1:4]	218 -144 350 416
H	num [1:4]	4.98e+11 -6.30e+10 5.29e...

Outline

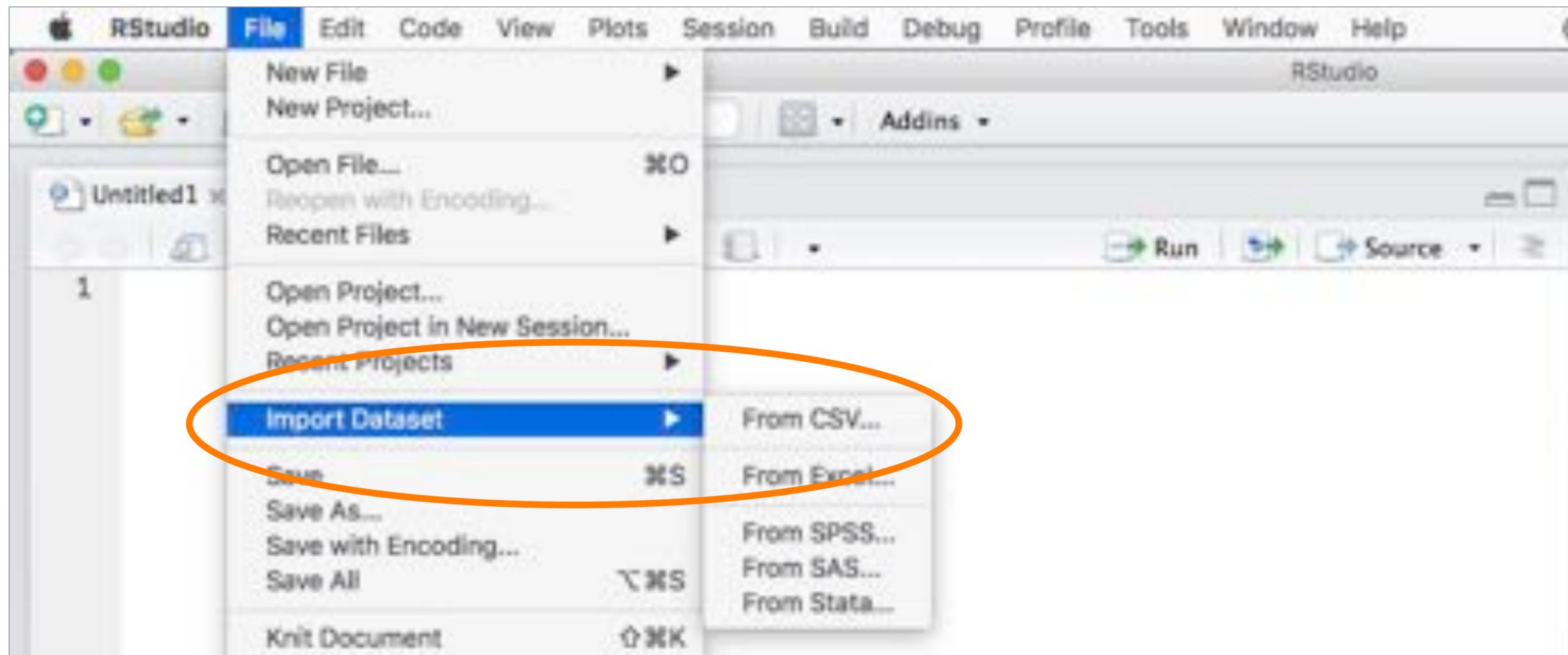
- What is data science?
- A data scientist's approach
- Introduction to R
 - Calculations in R
 - Reading data into R
 - Manipulating data in R
- Visualization in R
 - Basic plotting

A note about data

- Data can be found on a variety of sites on the internet
- Processing data stored in different formats is covered in a separate course
- For the purposes of this course, we will provide all the data sets already cleaned

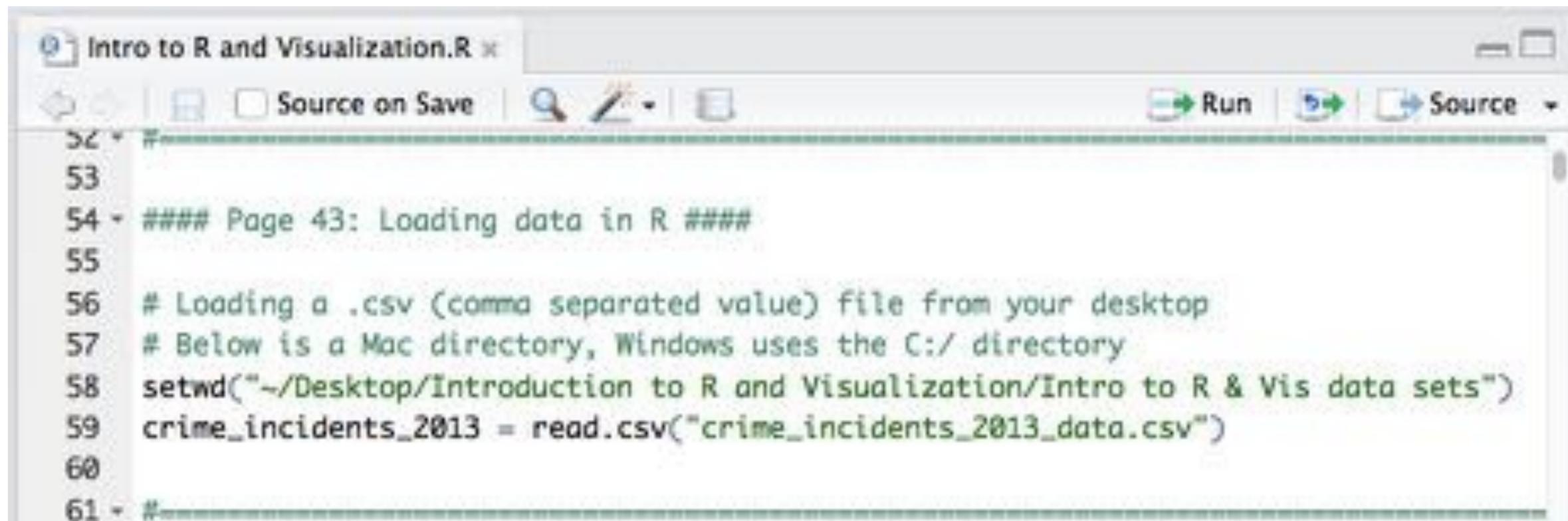
Loading data in R

- Loading data from your computer
 - Point and click



Loading data in R

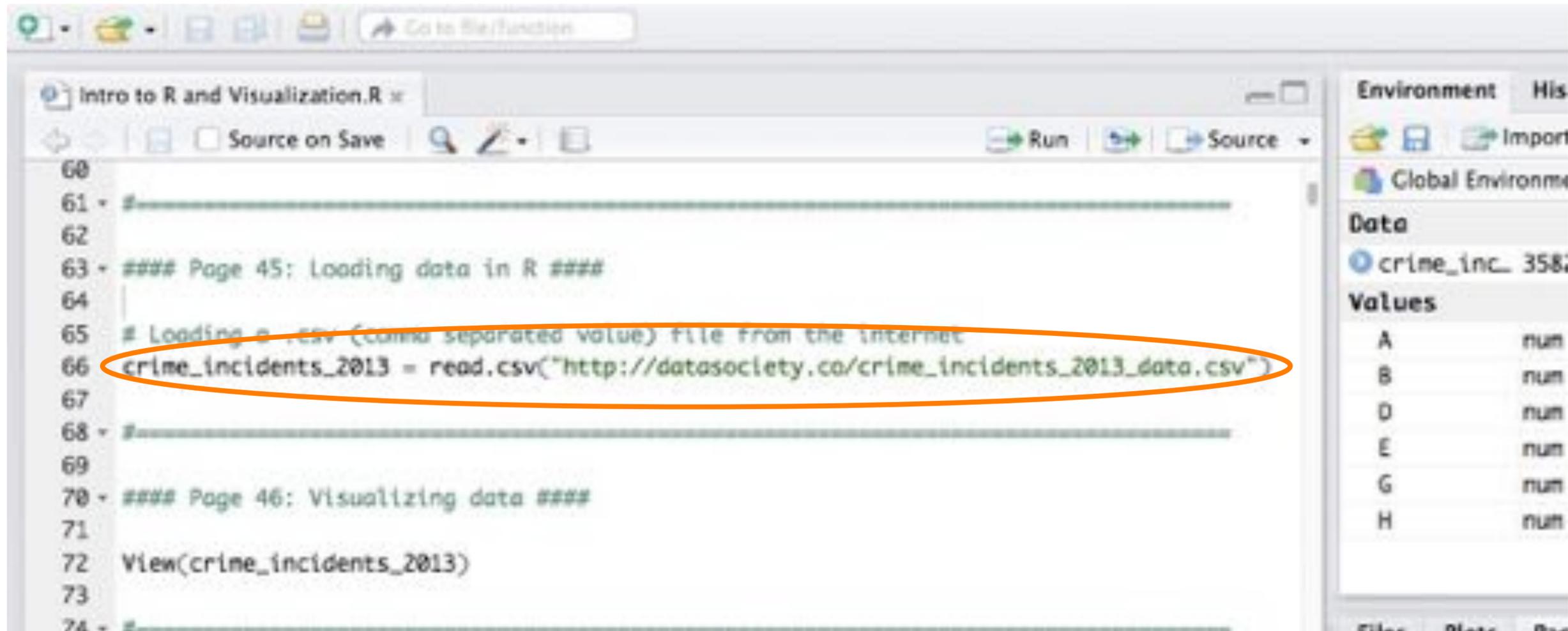
- Loading data from your computer
 - Enter code into script window
 - `crime_incidents_2013` is instantiated as the label of the data set



```
Intro to R and Visualization.R x
Source on Save
Run
Source
52 * #-----
53
54 * #### Page 43: Loading data in R ####
55
56 # Loading a .csv (comma separated value) file from your desktop
57 # Below is a Mac directory, Windows uses the C:/ directory
58 setwd("~/Desktop/Introduction to R and Visualization/Intro to R & Vis data sets")
59 crime_incidents_2013 = read.csv("crime_incidents_2013_data.csv")
60
61 * #-----
```

Loading data in R

- Loading data from the internet
 - Enter code into script window
 - `crime_incidents_2013` is instantiated as the label of the data set



The screenshot shows the R Studio interface. The script window on the left contains the following R code:

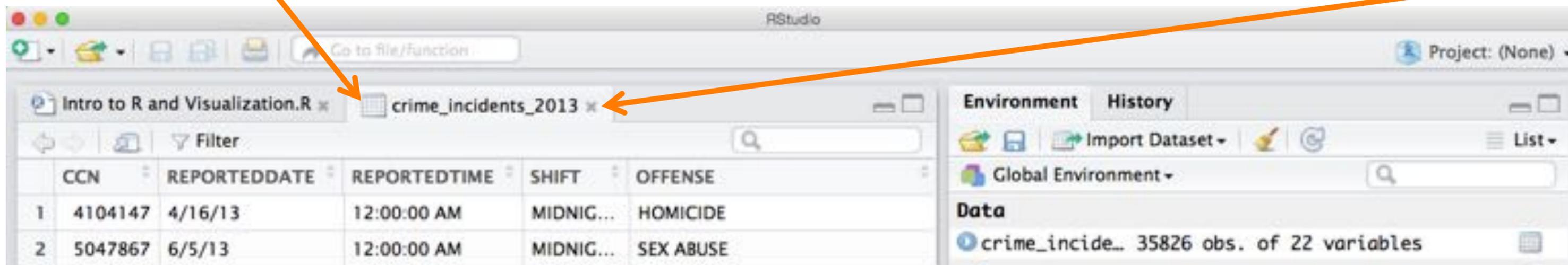
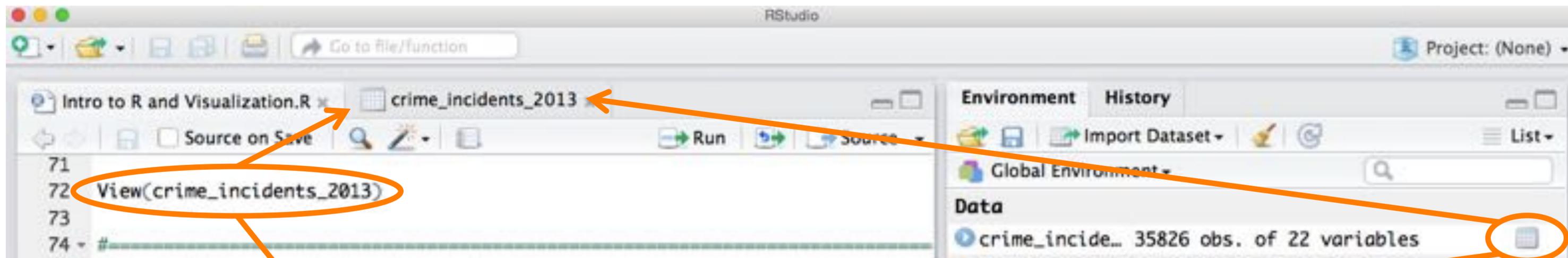
```
60  
61 #-----  
62  
63 ### Page 45: Loading data in R ###  
64  
65 # Loading a .csv (comma separated value) file from the internet  
66 crime_incidents_2013 = read.csv("http://datasociety.co/crime_incidents_2013_data.csv")  
67  
68 #-----  
69  
70 ### Page 46: Visualizing data ###  
71  
72 View(crime_incidents_2013)  
73  
74 #-----
```

The line `crime_incidents_2013 = read.csv("http://datasociety.co/crime_incidents_2013_data.csv")` is circled in orange. The environment window on the right shows the following:

Environment	
Global Environment	
Data	
crime_inc_358	
Values	
A	num
B	num
D	num
E	num
G	num
H	num

Visualizing data

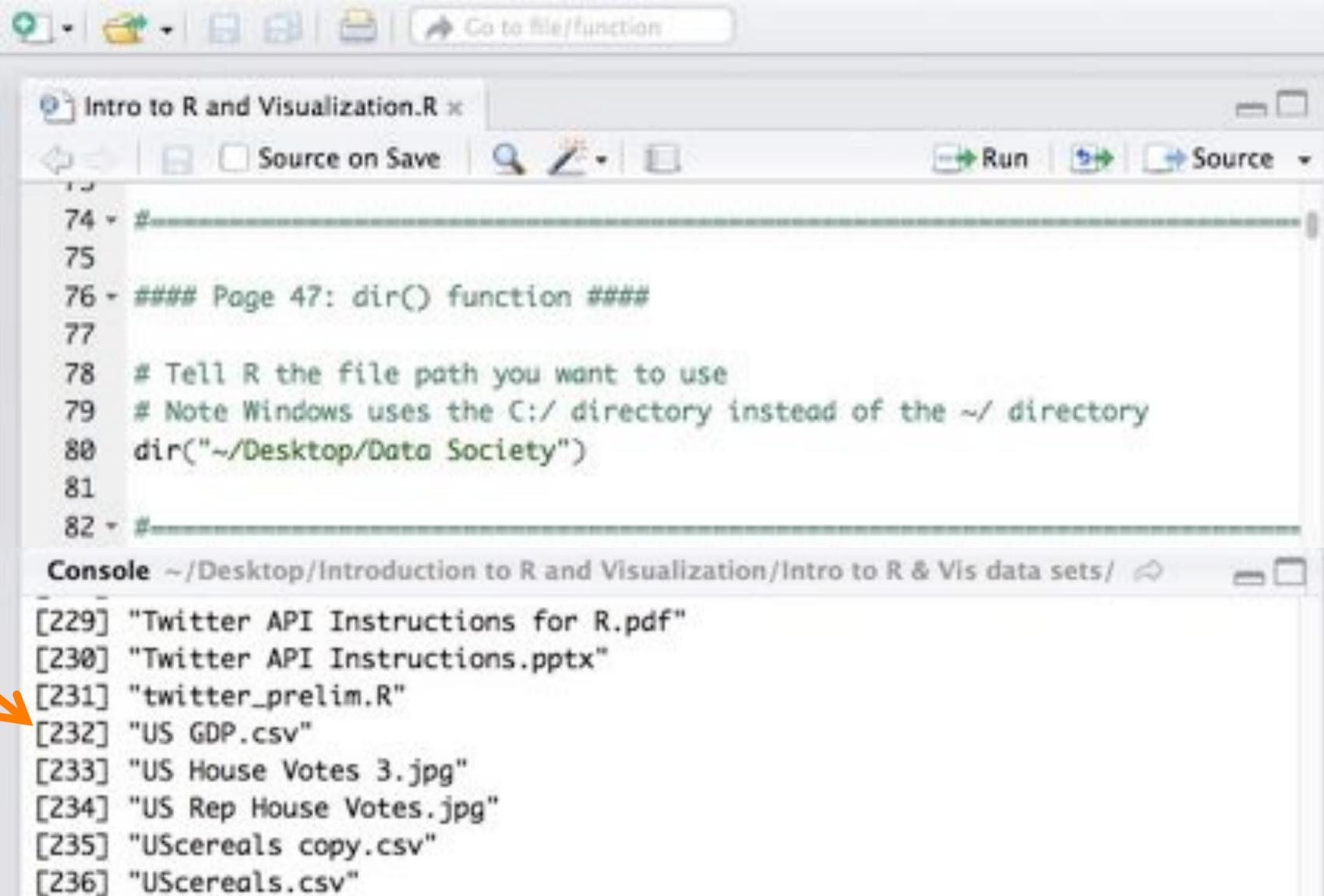
- Once data is loaded, you can see it as a spreadsheet by either:
 - Pressing the “spreadsheet” button in the top right window
 - Using the `View()` function in the script window



dir () function

- Lists all the files in a particular directory

Number of
the file in
the list



```
Intro to R and Visualization.R x
Source on Save Run Source
74 - #-----
75
76 - #### Page 47: dir() function ####
77
78 # Tell R the file path you want to use
79 # Note Windows uses the C:/ directory instead of the ~/ directory
80 dir("~/Desktop/Data Society")
81
82 - #-----

Console ~/Desktop/Introduction to R and Visualization/Intro to R & Vis data sets/
[229] "Twitter API Instructions for R.pdf"
[230] "Twitter API Instructions.pptx"
[231] "twitter_prelim.R"
[232] "US GDP.csv"
[233] "US House Votes 3.jpg"
[234] "US Rep House Votes.jpg"
[235] "UScereals copy.csv"
[236] "UScereals.csv"
```

R can read many types of files

Script

```
read.csv("filename.csv")      # read Excel files converted to csv format
read.table("filename")        # reads a table from a text file
read.spss("filename.spss")    # reads SPSS files
read.dta("filename.dta")      # reads Stata files
read.ssd("filename.ssd")      # reads SAS files
read.octave("filename.octave") # read Octave files
read.mtp("filename.mtp")      # read Minitab files
read.systat("filename.systat") # read Systat files
read.JPEG("filename.jpg")     # read JPEG image files
```

Note: this requires us to install package called 'jpeg', we will cover packages later

Types of data used in R

Basic (units)

- Integers (-1, 5, 100)
- Numerics (2.54)
- Characters ("Hello")
- Logicals (TRUE)
- Factors (A B C)

Composites

- Vectors
- Matrices (arrays)
- Lists
- Data frames

Basics make up composites

Note: R works only with types of data it "understands." When loading new data, you need to tell R what kind of data you are loading

Why do data types matter?

- Different functions and commands in R can only work with certain types of data
- You may need to tell R that data from a spreadsheet is either a vector a list or another type of data before you can perform analysis on it
- Note that a spreadsheet that only includes numbers can be called either a list or a matrix
- You should tell R how to interpret the data based on what analysis you'd like to perform

Vectors in R

Vectors allow you to automatically sort text documents, images, etc.

- A vector is a collection of elements of the same type (a column with either all numbers or all letters)
- R reads data as vectors
- Vectors allow you to manipulate a lot of data with a single command
- Operation between two vectors requires that vectors contain the same number of entries (same length)

Console

```
> # Create a vector
> x = c(1,2,3,4); x
[1] 1 2 3 4

> # Create a vector of years
> year = 2005:2010; year
[1] 2005 2006 2007 2008 2009 2010

> # Multiply constant by a vector
> x*3
[1] 3 6 9 12

> # Add two vectors
> y = c(1,2,2,1); x + y
[1] 2 4 5 5
```

";" separates commands that can be on separate lines

Vectors in R

A vector



	CCN	REPORTEDDATE	REPORTEDTIME	SHIFT	OFFENSE
1	4104147	4/16/13	12:00:00 AM	MIDNIC...	HOMICIDE
2	5047867	6/5/13	12:00:00 AM	MIDNIC...	SEX ABUSE
3	7083463	7/8/13	12:00:00 AM	MIDNIC...	SEX ABUSE
4	9172197	4/8/13	12:00:00 AM	MIDNIC...	SEX ABUSE
5	9251354	2/27/13	12:00:00 AM	MIDNIC...	SEX ABUSE
6	100289...	2/27/13	12:00:00 AM	MIDNIC...	SEX ABUSE
7	100335...	10/10/13	12:00:00 AM	MIDNIC...	SEX ABUSE
8	101249...	4/9/13	12:00:00 AM	MIDNIC...	SEX ABUSE
9	110101...	7/31/13	12:00:00 AM	MIDNIC...	HOMICIDE
10	110455...	1/31/13	12:00:00 AM	MIDNIC...	HOMICIDE
11	112502...	7/8/13	12:00:00 AM	MIDNIC...	SEX ABUSE

Showing 1 to 12 of 35,826 entries

Matrices in R

Matrices allow us to work with several columns of data at once

- A matrix is one or more vectors stacked next to each other
- Matrices can have row and column names, which can be determined and/or assigned (i.e. a customer list) by `dimnames`, `rownames` or `colnames` functions
- A matrix is a table where all the data is of the same type (i.e. numbers or letters)

Console

```
> # Create a matrix
> mat = matrix(1:6, nrow = 3, ncol = 2)
> mat
      [,1] [,2]
[1,]    1    4
[2,]    2    5
[3,]    3    6
```

Script

```
# Get matrix information
class(mat)          # "matrix"
is.vector(mat)      # FALSE
is.matrix(mat)      # TRUE
length(mat)         # 6
dim(mat)            # 3 2
```

Customizing matrices in R

Add names and adjust data order

- If you want to put in your data by row instead of by column, type `byrow = TRUE`
- You can add names for rows and columns by using the function `dimnames`

```
> mat1 = matrix(1:6, nrow = 3, ncol = 2, Console
byrow = TRUE)
> mat1
      [,1] [,2]
[1,]    1    2
[2,]    3    4
[3,]    5    6

> mat2 = matrix(1:6, nrow = 3, ncol = 2,
byrow = TRUE, dimnames = list(c("Row1",
"Row2", "Row3"), c("Col1", "Col2")))
> mat2
      Col1 Col2
Row1     1    2
Row2     3    4
Row3     5    6
```

Lists in R

Lists allow you to work with different types of data mixed together

- A list is a vector with different types of elements (data)
- The elements of a list can be numeric vectors, character vectors, matrices and other lists
- List components are determined by \$ signs

Console

```
> # Create a list
> Johnsons = list(husband = "Bill", wife =
"Joanna", children = TRUE, child.ages = c(3,
13, 18))
> Johnsons
$husband
[1] "Bill"

$wife
[1] "Joanna"

$children
[1] TRUE

$child.ages
[1] 3 13 18
```

Lists in R

You can pull out detailed information from lists

- `length("list")`: gives you the number of components
- `class("list")`: identifies the data type of the information
- `names("list")`: identifies the name of each component
- `"list"[1:2]`: identifies the specific data in those components
- `"list"$"component_name"`: identifies the specific data in that component

```
> # Get list information
> length(Johnsons)
[1] 4
> class(Johnsons)
[1] "list"
> names(Johnsons)
[1] "husband"      "wife"          "children"
[4] "child.ages"
> Johnsons[1:2]
$husband
[1] "Bill"

$wife
[1] "Joanna"

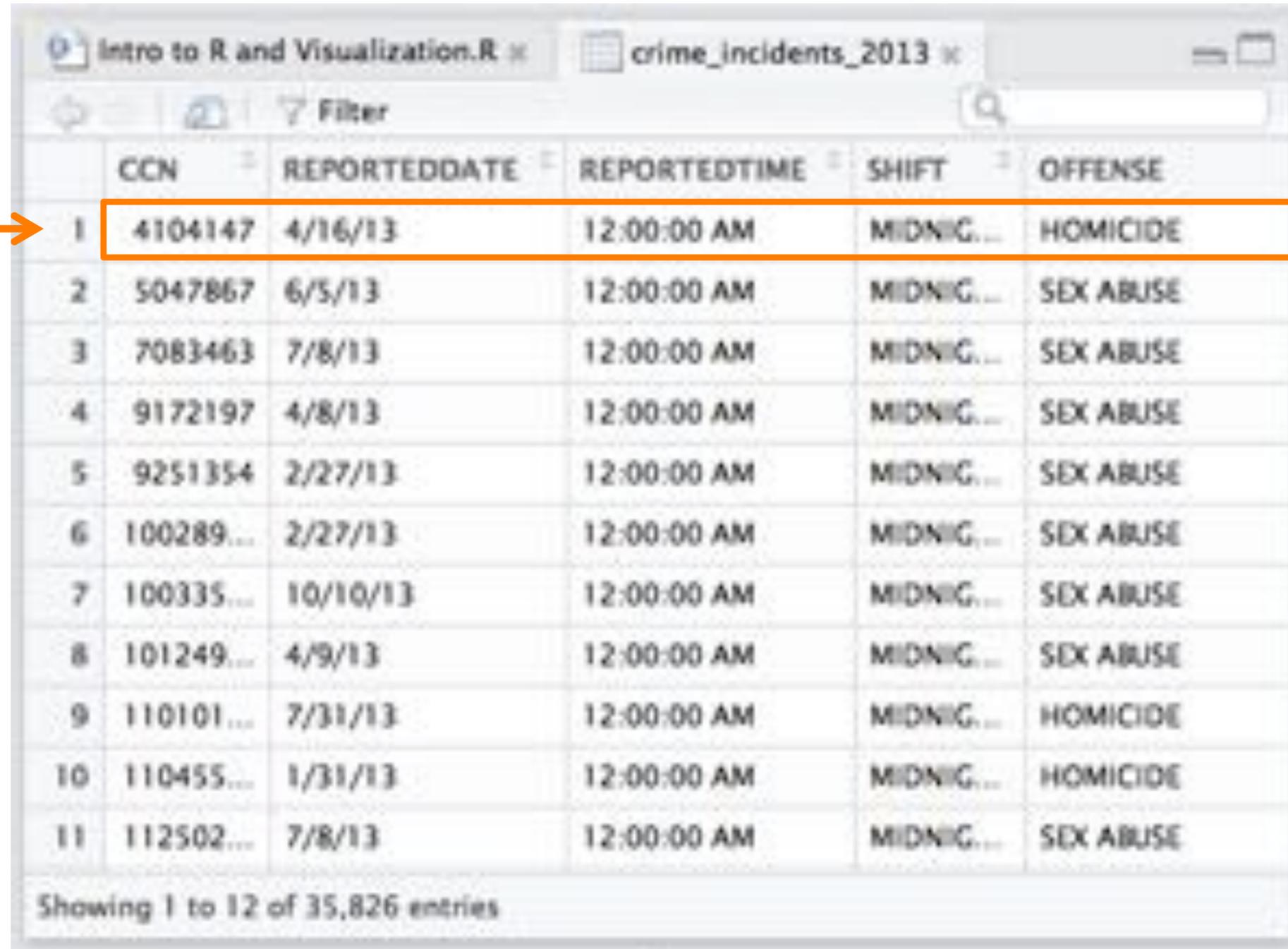
> Johnsons$child.ages
[1] 3 13 18
```

Console

R automatically numbers the components

Lists in R

A list includes several data types and unstructured data



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1	4104147	4/16/13	12:00:00 AM	MIDNIG...	HOMICIDE
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4	9172197	4/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE
5	9251354	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE
6	100289...	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE
7	100335...	10/10/13	12:00:00 AM	MIDNIG...	SEX ABUSE
8	101249...	4/9/13	12:00:00 AM	MIDNIG...	SEX ABUSE
9	110101...	7/31/13	12:00:00 AM	MIDNIG...	HOMICIDE
10	110455...	1/31/13	12:00:00 AM	MIDNIG...	HOMICIDE
11	112502...	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE

Showing 1 to 12 of 35,826 entries

Data frames in R

Data frames allow you to work with alpha-numeric and other data types

- Data frames are matrices with different data types
- Each column is a vector of the same length, types may differ
- The elements of a data frame can be numeric vectors, factor vectors, and logical vectors
- Consist of observations (rows), variables (columns)

Console

```
> # Setting up variables for data.frame
> first.name = c('Joe', 'Bob', 'Jill')
> last.name = c('Li', 'Dan', 'Smith')
> age = c(45, 20, 37)
```

```
> # Setting up data.frame
> df = data.frame(first.name,
> last.name, age)
```

```
> # Output
```

```
> df
```

	first.name	last.name	age
1	Joe	Li	45
2	Bob	Dan	20
3	Jill	Smith	37

Combining lists into data frames

You can combine lists and create data frames

- Note that these lists must have the same number of components
- Use function `rbind` (row bind) to match the categories and combine lists
 - When using `rbind`, the names of the columns must be the same

Note: the `rbind` function won't work with multiple data points within an element of a list

Script

```
# Combining lists into a data frame

Simpsons = list(husband = "Homer", wife =
                "Marge", children = TRUE,
                num.kids = 3)

Belchers = list(husband = "Bob", wife =
                "Linda", children = TRUE,
                num.kids = 3)

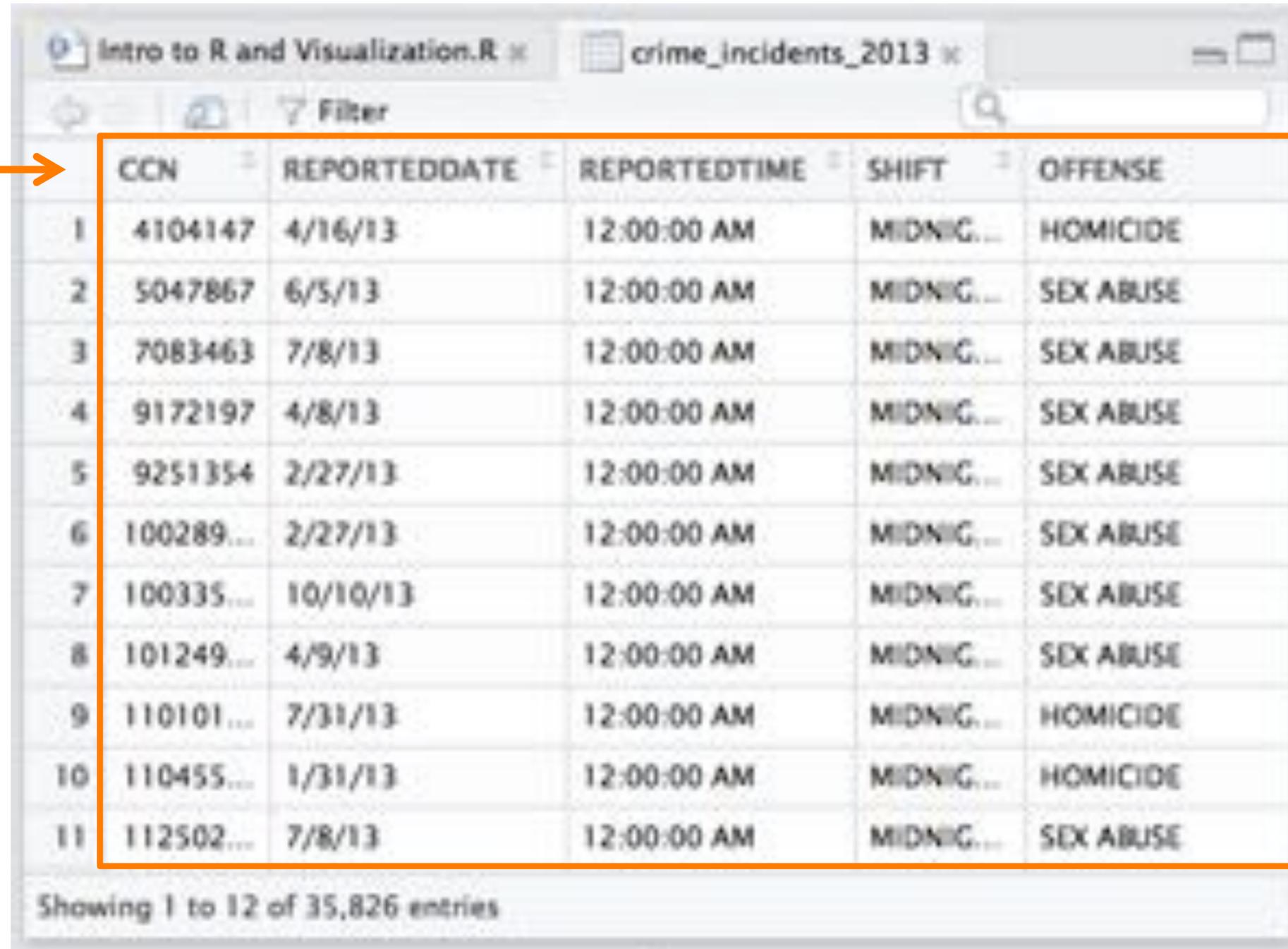
rbind(Simpsons, Johnsons)
```

Console

```
> rbind(Simpsons, Belchers)
      husband wife  children num.kids
Simpsons "Homer" "Marge"  TRUE      3
Belchers "Bob"   "Linda"  TRUE      3
```

Data frames in R

A data frame includes several data types



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7	100335...	10/10/13	12:00:00 AM	MIDNIG...	SEX ABUSE
8	101249...	4/9/13	12:00:00 AM	MIDNIG...	SEX ABUSE
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Showing 1 to 12 of 35,826 entries

Creating data types in R

Composite data types	Instantiating (creating)	Appropriate use
Vector	<code>c(. . .)</code>	Store simple data types in a row or column
Matrix	<code>matrix(data, nrows, ncols)</code>	Store multiple rows and columns of a single data type
Data Frame	<code>data.frame(data matrix)</code>	Work with alphanumeric and other data types
List	<code>list(element_1, ..., element_k)</code>	Work with different data types simultaneously

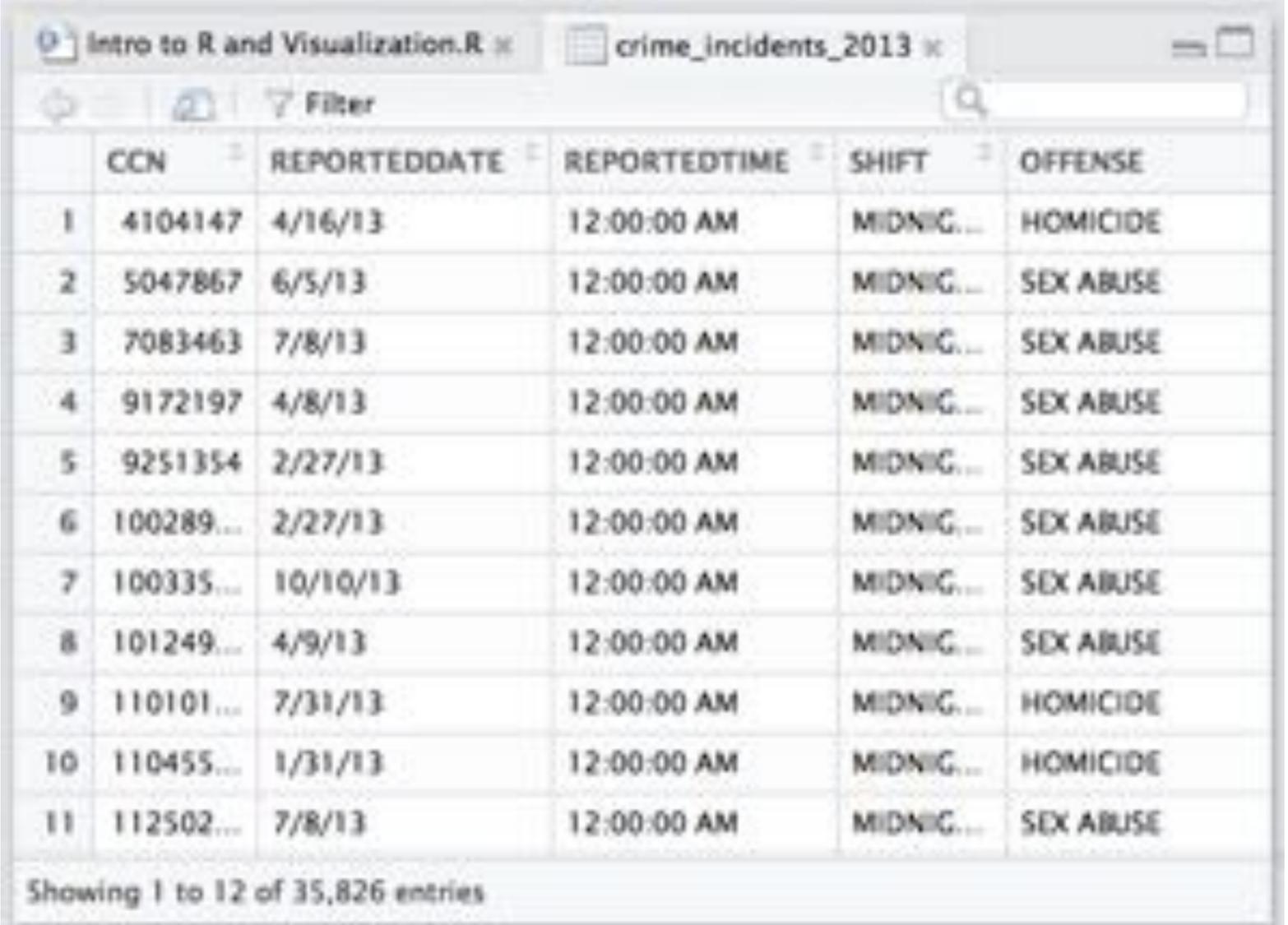
Outline

- What is data science?
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DC crime data 2013

What does the data look like?

- Almost 36,000 rows, each row represents a crime
- Information includes date, time, offense, method, location, precinct, etc.
- Always take the time to look over your data – this habit will help you understand the information you are working with



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4	9172197	4/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE
5	9251354	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE
6	100289...	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE
7	100335...	10/10/13	12:00:00 AM	MIDNIG...	SEX ABUSE
8	101249...	4/9/13	12:00:00 AM	MIDNIG...	SEX ABUSE
9	110101...	7/31/13	12:00:00 AM	MIDNIG...	HOMICIDE
10	110455...	1/31/13	12:00:00 AM	MIDNIG...	HOMICIDE
11	112502...	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE

Showing 1 to 12 of 35,826 entries

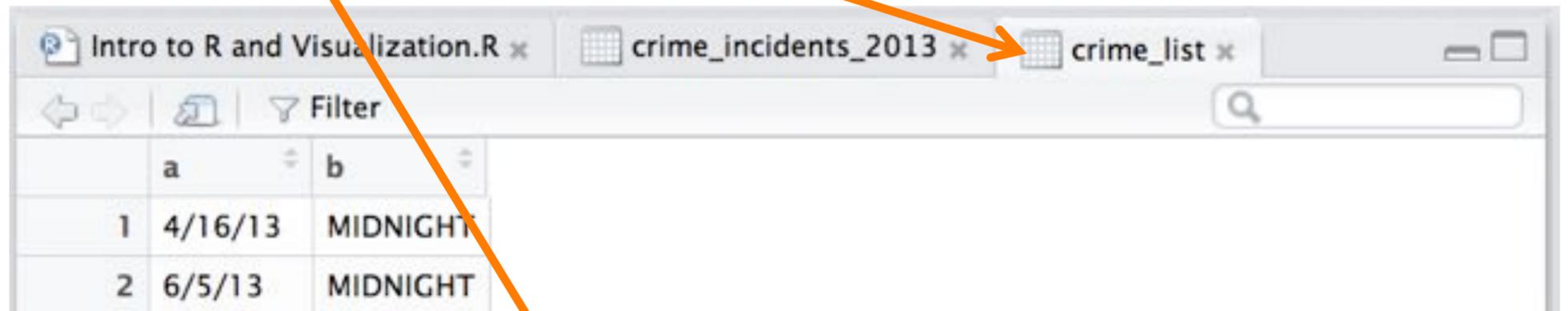
Subsetting data

Denotes what data set to use Denotes what column to use

- You can separate out certain columns of the data by using a `$` sign
- The format is `"data_set" $"name_of_column"`
- If the list you are creating has **columns of different lengths, the `View()` function will not work**
 - Use the `head()` function or just enter the new variable you created instead

```
Script  
crime_list = list(a = crime_incidents_2013$REPORTEDDATE,  
                 b = crime_incidents_2013$SHIFT)  
  
View(crime_list)  
head(crime_list)  
crime_list
```

Tells R that we are using a piece of the data set



	a	b
1	4/16/13	MIDNIGHT
2	6/5/13	MIDNIGHT

```
Console ~/Dropbox/  
[9993] EVENING EVENING EVENING EVENING EVENING EVENING EVENING EVENING  
[ reached getOption("max.print") -- omitted 25826 entries ]  
Levels: DAY EVENING MIDNIGHT
```

Factors in R

- A factor is a unique value in a data set
- For example, if five crimes are labeled as “arson, burglary, burglary, burglary, robbery”, there would be three factors: arson, burglary, and robbery
- Use the `as.factor()` function to tell R to read data as a factor

```
# Set a category of our crime data set as a factor using as.factor()
crime_categories = as.factor(crime_incidents_2013$OFFENSE)

# Visualize the unique values in the OFFENSE category of the crime data set
levels(crime_categories)
View(levels(crime_categories))
```

Script

Factors in R

The screenshot displays the RStudio interface. In the top-left pane, the 'levels(crime_categories)' window shows a list of 9 crime categories: ARSON, ASSAULT W/DANGEROUS WEAPON, BURGLARY, HOMICIDE, MOTOR VEHICLE THEFT, ROBBERY, SEX ABUSE, THEFT F/AUTO, and THEFT/OTHER. The top-right pane shows the Environment window with 'crime_categories' listed as a 'Factor w/ 9 levels'. The bottom-left pane shows the Console with the following R code and output:

```
> crime_categories = as.factor(crime_incidents_2013$OFFENSE)
> levels(crime_categories)
[1] "ARSON" "ASSAULT W/DANGEROUS WEAPON"
[3] "BURGLARY" "HOMICIDE"
[5] "MOTOR VEHICLE THEFT" "ROBBERY"
[7] "SEX ABUSE" "THEFT F/AUTO"
[9] "THEFT/OTHER"
> View(levels(crime_categories))
```

The bottom-right pane shows the Files window with a directory listing:

Name	Size	Modified
.gitignore	12 B	Feb 22, 2015, 9:31 AM
.R		
.RData	42 B	Jan 25, 2015, 1:15 PM
.Rhistory	63 B	Aug 21, 2015, 7:13 PM
Applications		
Desktop		
Documents		
Downloads		
Dropbox		

Searching for terms in your data

- The function `grep()` can help you check if your data set includes a given term

```
# grep() only works on a data type called a character vector
# Recall that a vector is a single column
crime = as.character(crime_incidents_2013$METHOD)  Select the "METHOD" column

# Create a new variable that includes the output of grep()
gun_rows = grep("GUN", crime)

# View gun_rows, this includes all the row numbers with the word "GUN"
# Recall that the View() function displays data frames so tell R that "gun_rows"
# is a data frame
View(as.data.frame(gun_rows))
```

Script

- You can search the entire data set by combining all the columns into one or searching each column individually

Searching for terms in your data

- How many rows include “GUN” as the method?

```
> # Find the number of rows  
> length(gun_rows)  
[1] 2156
```

Console

Summarizing your data

- The `table()` function is a convenient way to see unique categories in your data and determine the frequency of occurrence

```
crime_summary = table(crime_incidents_2013$OFFENSE)
```

```
View(crime_summary)
```

Script

Column to summarize



	Var1	Freq
1	ARSON	35
2	ASSAULT W/DANGEROUS WEAPON	2387
3	BURGLARY	3365
4	HOMICIDE	103
5	MOTOR VEHICLE THEFT	2671
6	ROBBERY	4072
7	SEX ABUSE	298
8	THEFT F/AUTO	10101
9	THEFT/OTHER	12874

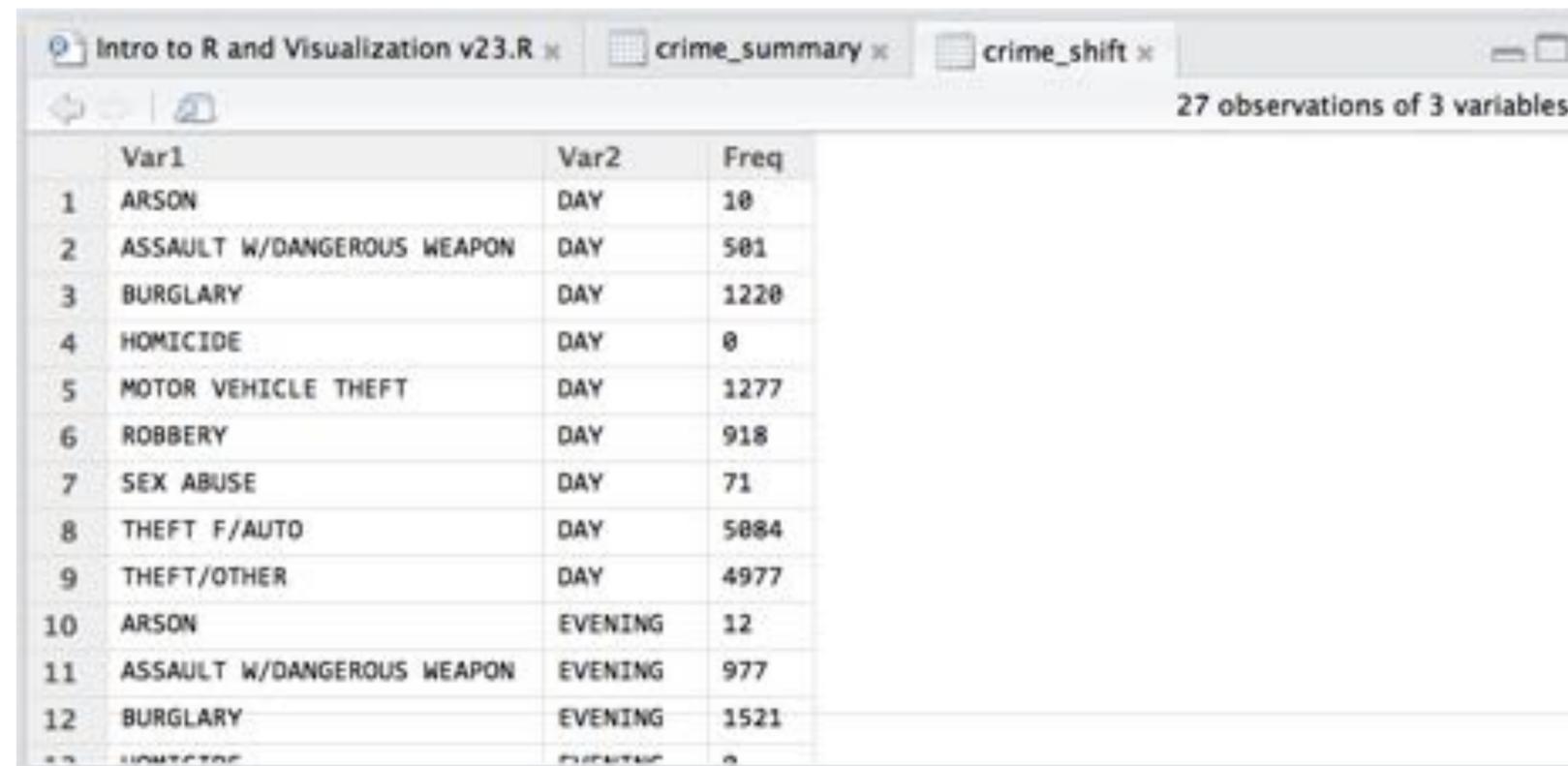
Summarizing your data

- What if I want to see how crime is distributed over 24 hours?

```
crime_time = table(crime_incidents_2013$OFFENSE,  
                  crime_incidents_2013$SHIFT)
```

```
View(crime_time)
```

Script



The screenshot shows an R Studio window with a table titled 'crime_summary'. The table has 27 observations and 3 variables: Var1 (Offense), Var2 (Shift), and Freq (Frequency). The data is as follows:

	Var1	Var2	Freq
1	ARSON	DAY	10
2	ASSAULT W/DANGEROUS WEAPON	DAY	501
3	BURGLARY	DAY	1220
4	HOMICIDE	DAY	0
5	MOTOR VEHICLE THEFT	DAY	1277
6	ROBBERY	DAY	918
7	SEX ABUSE	DAY	71
8	THEFT F/AUTO	DAY	5084
9	THEFT/OTHER	DAY	4977
10	ARSON	EVENING	12
11	ASSAULT W/DANGEROUS WEAPON	EVENING	977
12	BURGLARY	EVENING	1521

Sorting your data

- The `order()` function is a convenient way to sort your data
 - For more details of what this function can do, run the `?order` command

Create new data set

Data set used

Column to sort by

```
crime_order = crime_incidents_2013[order(crime_incidents_2013$METHOD,  
decreasing = FALSE),]  
View(crime_order)
```

Script

- Note that the `decreasing` operator can be set to `TRUE` or `FALSE` depending on the order you are looking for (i.e. ascending or descending)

Sorting your data

The screenshot shows an R Studio window with three tabs: 'Intro to R and Visualization.R', 'crime_incidents_2013', and 'crime_order'. The 'crime_order' tab is active, displaying a data table. The table has columns: CCN, REPORTEDDATE, REPORTEDTIME, SHIFT, OFFENSE, and METHOD. The METHOD column is highlighted with an orange border. The data is sorted by the METHOD column, with all entries being 'GUN'. The table shows 12 rows of data, with a status bar at the bottom indicating 'Showing 1 to 12 of 35,826 entries'.

	CCN	REPORTEDDATE	REPORTEDTIME	SHIFT	OFFENSE	METHOD
10	110455...	1/31/13	12:00:00 AM	MIDNIG...	HOMICIDE	GUN
19	121824...	1/1/13	12:15:00 AM	MIDNIG...	ROBBERY	GUN
21	121824...	1/1/13	12:18:00 AM	MIDNIG...	ROBBERY	GUN
25	121825...	1/1/13	2:29:00 AM	MIDNIG...	ROBBERY	GUN
28	121825...	1/1/13	3:12:00 AM	MIDNIG...	ASSAULT W/DANGEROUS WEAP...	GUN
32	121825...	1/1/13	4:23:00 AM	MIDNIG...	ASSAULT W/DANGEROUS WEAP...	GUN
37	121826...	1/1/13	7:30:00 AM	DAY	ROBBERY	GUN
67	130001...	1/1/13	4:49:00 PM	EVENING	ROBBERY	GUN
69	130001...	1/1/13	7:00:00 PM	EVENING	ROBBERY	GUN
73	130002...	1/1/13	7:56:00 PM	EVENING	BURGLARY	GUN
118	130004...	1/2/13	11:39:00 AM	DAY	ROBBERY	GUN

Showing 1 to 12 of 35,826 entries

Eliminating duplicate records

- You can eliminate repetitive instances of your data using `unique()`
 - Eliminates rows where every entry is the same as another row

```
crime_no_dups = unique(crime_incidents_2013)

# Check how many rows are identical by looking at the dimensions of each data set
dim(crime_incidents_2013)
dim(crime_no_dups)

# You can just compute the difference in the dimensions of the data set
dim(crime_incidents_2013) - dim(crime_no_dups)
```

Script

9 rows
were
eliminated

```
Console ~/Desktop/Course Revisions/Week 1/
> dim(crime_incidents_2013)
[1] 35826  22
> dim(crime_no_dups)
[1] 35817  22
> dim(crime_incidents_2013) - dim(crime_no_dups)
[1] 9
```

Subsetting: real data example

The screenshot shows the RStudio interface with a data table loaded. The table has 7 columns: CCN, REPORTEDDATE, REPORTEDTIME, SHIFT, OFFENSE, METHOD, and LASTMODIFIEDDATE. The 'REPORTEDDATE' column is highlighted with a yellow circle. The 'DAY' value in the 13th row is highlighted with a blue circle. The table shows 15 rows of data, with a total of 35,826 entries.

	CCN	REPORTEDDATE	REPORTEDTIME	SHIFT	OFFENSE	METHOD	LASTMODIFIEDDATE
1	4104147	4/16/13	12:00:00 AM	MIDNIG...	HOMICIDE	KNIFE	6/23/14
2	5047867	6/5/13	12:00:00 AM	MIDNIG...	SEX ABUSE	KNIFE	6/23/14
3	7083463	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
4	9172197	4/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
5	9251354	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
6	100289...	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
7	100335...	10/10/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
8	101249...	4/9/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
9	110101...	7/31/13	12:00:00 AM	MIDNIG...	HOMICIDE	OTHERS	6/23/14
10	110455...	1/31/13	12:00:00 AM	MIDNIG...	HOMICIDE	GUN	6/23/14
11	112502...	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
12	120035...	1/9/13	12:59:00 AM	MIDNIG...	THEFT/OTHER	OTHERS	6/23/14
13	120375...	3/23/13	10:00:00 AM	DAY	THEFT/OTHER	OTHERS	6/23/14
14	120557...	8/19/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
15	120581...	5/13/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14

Showing 1 to 15 of 35,826 entries

Subsetting: selects only values you want

- When working with large data sets, you may want to use or visualize only a portion of your data at any given time

```
# Vector subsetting
v = c(10, 20, 30, 40, 50, 60) # define your vector with 6 numbers
v[3] # select the 3rd term
v[1:4] # select the first 4 terms
v[3:6] # select the 3rd through the 6th term
v[c(1, 3, 6)] # select the 1st, 3rd and 6th term
2*v[3:6] # multiply the 3rd through the 6th term by 2
```

```
[1] 30
[1] 10 20 30 40
[1] 30 40 50 60
[1] 10 30 60
[1] 60 80 100 120
```

Subsetting: selects only values you want

- When working with large data sets, you may want to use or visualize only a portion of your data at any given time

```
# Matrix subsetting
m = matrix(c(5, 6, 7, 8), 2, 2) # define your matrix with 4 numbers in
                                2 rows and 2 columns

m                                # view your matrix
m[2, ]                          # select the numbers in the 2nd row
m[, 1]                          # select the numbers in the 1st column
m[1, 2]                         # select the number in the 1st row and 2nd column
m[1:3]                          # select the first 3 terms in the matrix
m[1:2, 1:2]                    # select the #s in 1st and 2nd rows and 1st and 2nd cols
2*m[1:3]                       # multiply the first 3 terms in the matrix by 2
```

Script

Subsetting: selects only values you want

Console

```
> m # view your matrix
      [,1] [,2]
[1,]    5    7
[2,]    6    8
> m[2,] # select the numbers in the 2nd row
[1] 6 8
> m[,1] # select the numbers in the 1st column
[1] 5 6
> m[1,2] # select the number in the 1st row and 2nd column
[1] 7
> m[1:3] # select the first 3 terms in the matrix
[1] 5 6 7
> m[1:2,1:2] # select the #s in 1st and 2nd rows and 1st and 2nd cols
      [,1] [,2]
[1,]    5    7
[2,]    6    8
> 2*m[1:3] # multiply the first 3 terms in the matrix by 2
[1] 10 12 14
```

Subsetting: real data example

The screenshot shows the RStudio interface with a data table loaded. The table has the following columns: CCN, REPORTEDDATE, REPORTEDTIME, SHIFT, OFFENSE, METHOD, and LASTMODIFIEDDATE. The data is displayed in a grid format. The REPORTEDDATE column is highlighted in yellow, and the DAY column in row 13 is highlighted in blue. The text 'Showing 1 to 15 of 35,826 entries' is visible at the bottom of the table.

	CCN	REPORTEDDATE	REPORTEDTIME	SHIFT	OFFENSE	METHOD	LASTMODIFIEDDATE
1	4104147	4/16/13	12:00:00 AM	MIDNIG...	HOMICIDE	KNIFE	6/23/14
2	5047867	6/5/13	12:00:00 AM	MIDNIG...	SEX ABUSE	KNIFE	6/23/14
3	7083463	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
4	9172197	4/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
5	9251354	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
6	100289...	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
7	100335...	10/10/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
8	101249...	4/9/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
9	110101...	7/31/13	12:00:00 AM	MIDNIG...	HOMICIDE	OTHERS	6/23/14
10	110455...	1/31/13	12:00:00 AM	MIDNIG...	HOMICIDE	GUN	6/23/14
11	112502...	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
12	120035...	1/9/13	12:59:00 AM	MIDNIG...	THEFT/OTHER	OTHERS	6/23/14
13	120375...	3/23/13	10:00:00 AM	DAY	THEFT/OTHER	OTHERS	6/23/14
14	120557...	8/19/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14
15	120581...	5/13/13	12:00:00 AM	MIDNIG...	SEX ABUSE	OTHERS	6/23/14

Showing 1 to 15 of 35,826 entries

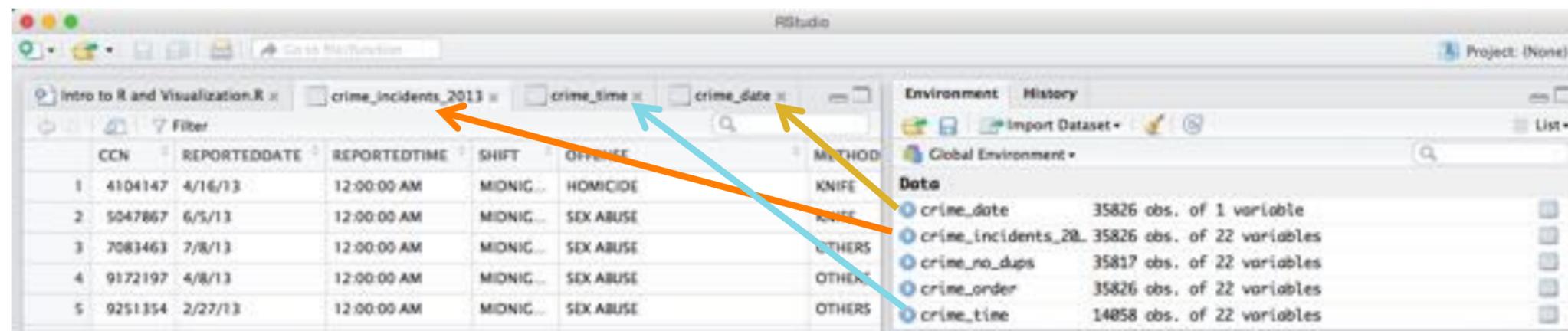
Subsetting: real data example

Script

```
# Select data from the crime data set based on time of day the crime occurred  
crime_time = subset(crime_incidents_2013, SHIFT == "DAY")
```

```
# Or you can also use  
crime_time = crime_incidents_2013[crime_incidents_2013$SHIFT == "DAY",]  
View(crime_time)
```

```
# Select 1 column from the data set  
crime_date = subset(crime_incidents_2013, select = REPORTEDDATE)  
View(crime_date)
```



The screenshot shows the RStudio interface. The Environment pane on the right lists the following objects:

Object	Details
Global Environment	
crime_date	35826 obs. of 1 variable
crime_incidents_2013	35826 obs. of 22 variables
crime_no_dups	35817 obs. of 22 variables
crime_order	35826 obs. of 22 variables
crime_time	14858 obs. of 22 variables

The Data Viewer on the left shows a table with the following columns: CCN, REPORTEDDATE, REPORTEDTIME, SHIFT, OFFENSE, and METHOD. The first five rows of data are:

CCN	REPORTEDDATE	REPORTEDTIME	SHIFT	OFFENSE	METHOD
1	4104147	4/16/13	12:00:00 AM	MIDNIG...	HOMICIDE
2	5047867	6/5/13	12:00:00 AM	MIDNIG...	SEX ABUSE
3	7083463	7/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE
4	9172197	4/8/13	12:00:00 AM	MIDNIG...	SEX ABUSE
5	9251354	2/27/13	12:00:00 AM	MIDNIG...	SEX ABUSE

Arrows in the image point from the Environment pane to the Data Viewer, indicating the relationship between the objects and the data shown.

Subsetting: real data example

RStudio interface showing the initial data table. The variable `crime_incidents_2013` is highlighted in orange in the top toolbar. The Environment pane on the right shows the following data objects:

Variable	Observations	Variables
<code>crime_date</code>	35826	1
<code>crime_incidents_2013</code>	35826	22
<code>crime_no_dups</code>	35817	22
<code>crime_order</code>	35826	22
<code>crime_time</code>	14858	22

RStudio interface showing the data table filtered by `crime_time`, which is highlighted in blue in the top toolbar. The Environment pane on the right shows the same data objects as in the first screenshot.

RStudio interface showing the data table filtered by `crime_date`, which is highlighted in yellow in the top toolbar. The Environment pane on the right shows the same data objects as in the previous screenshots.

Subsetting: real data example

- What if we wanted to select only robberies that happened during the day?
 - When performing several operations, keep the steps separate so it's easier to check your code

```
# First, only select the crimes that happened during the day
crime_time = subset(crime_incidents_2013, SHIFT == "DAY")

# Second, select only robberies out of the new data set called "crime_time"
crime_time_robbery = subset(crime_time, OFFENSE == "ROBBERY")
View(crime_time_robbery)
```

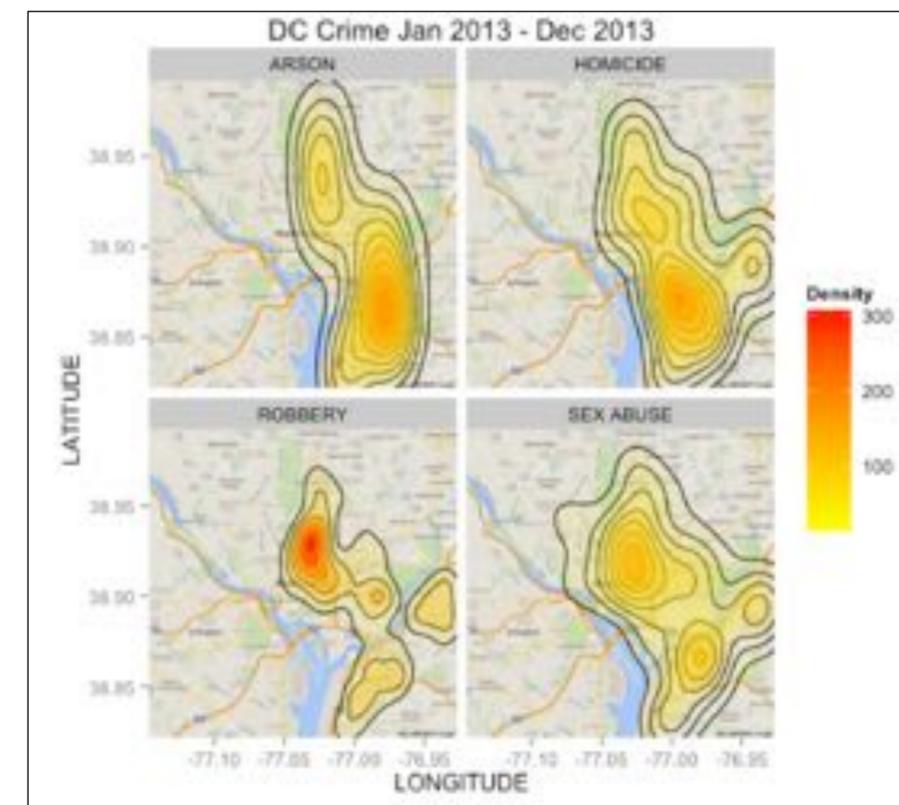
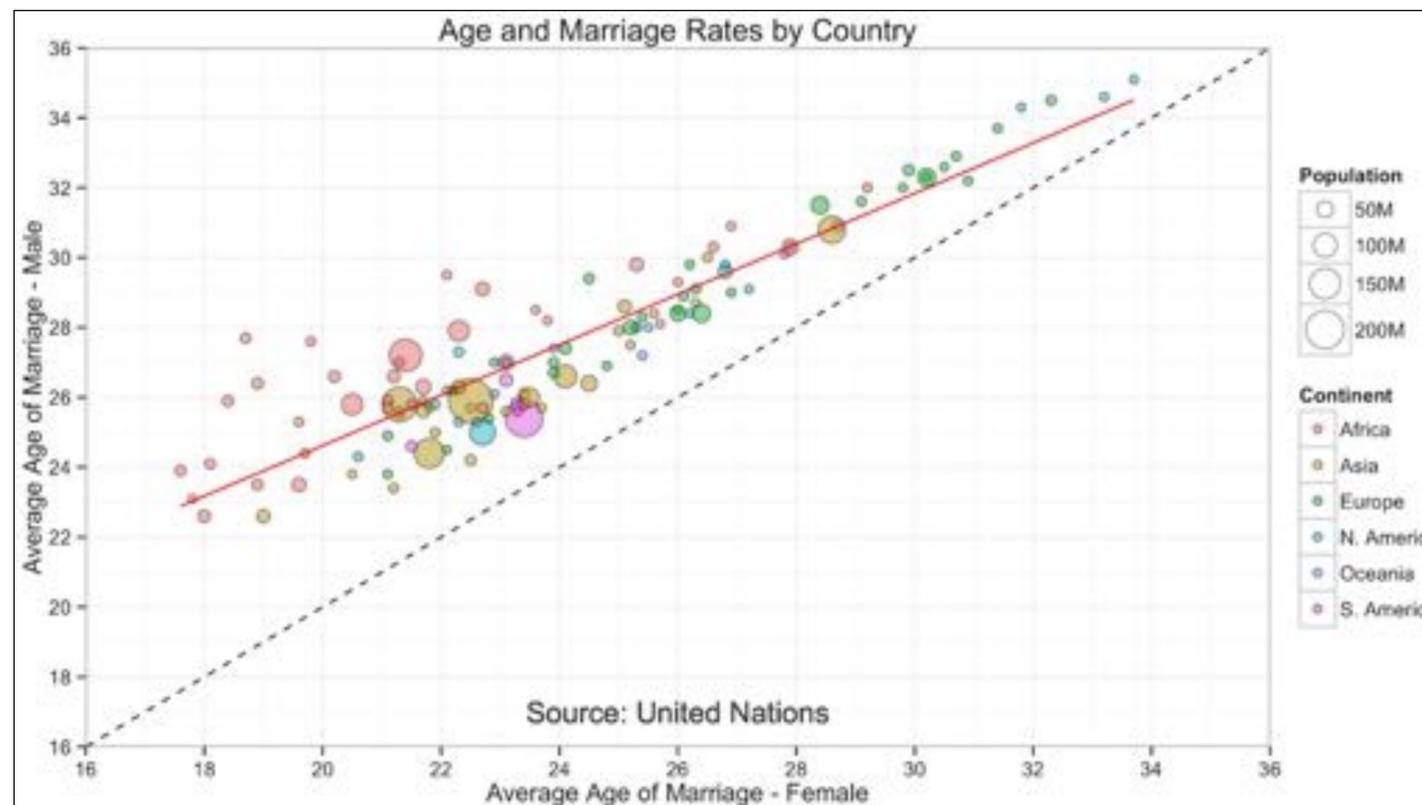
Script

Outline

- What is data science?
- A data scientist's approach
- Introduction to R
 - Calculations in R
 - Reading data into R
 - Manipulating data in R
- Visualization in R
 - Basic plotting

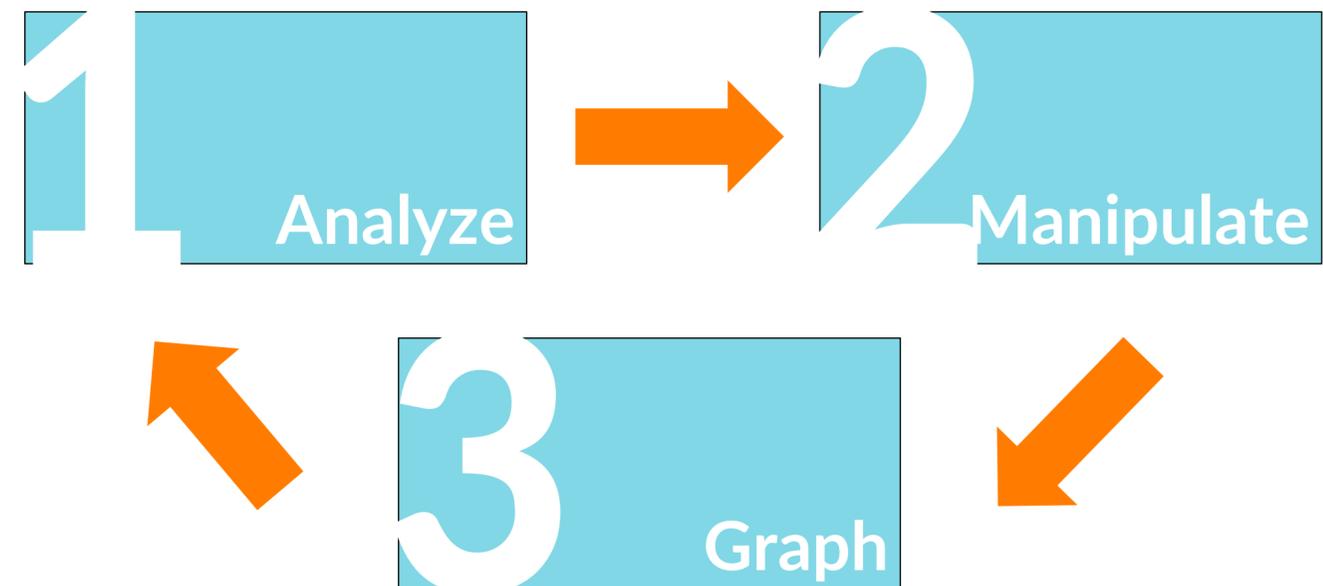
Why build a visualization?

- To communicate your ideas
- To better understand your data
- To discover a new insight
- Visualization is a great way to do exploratory data analysis (“EDA”)



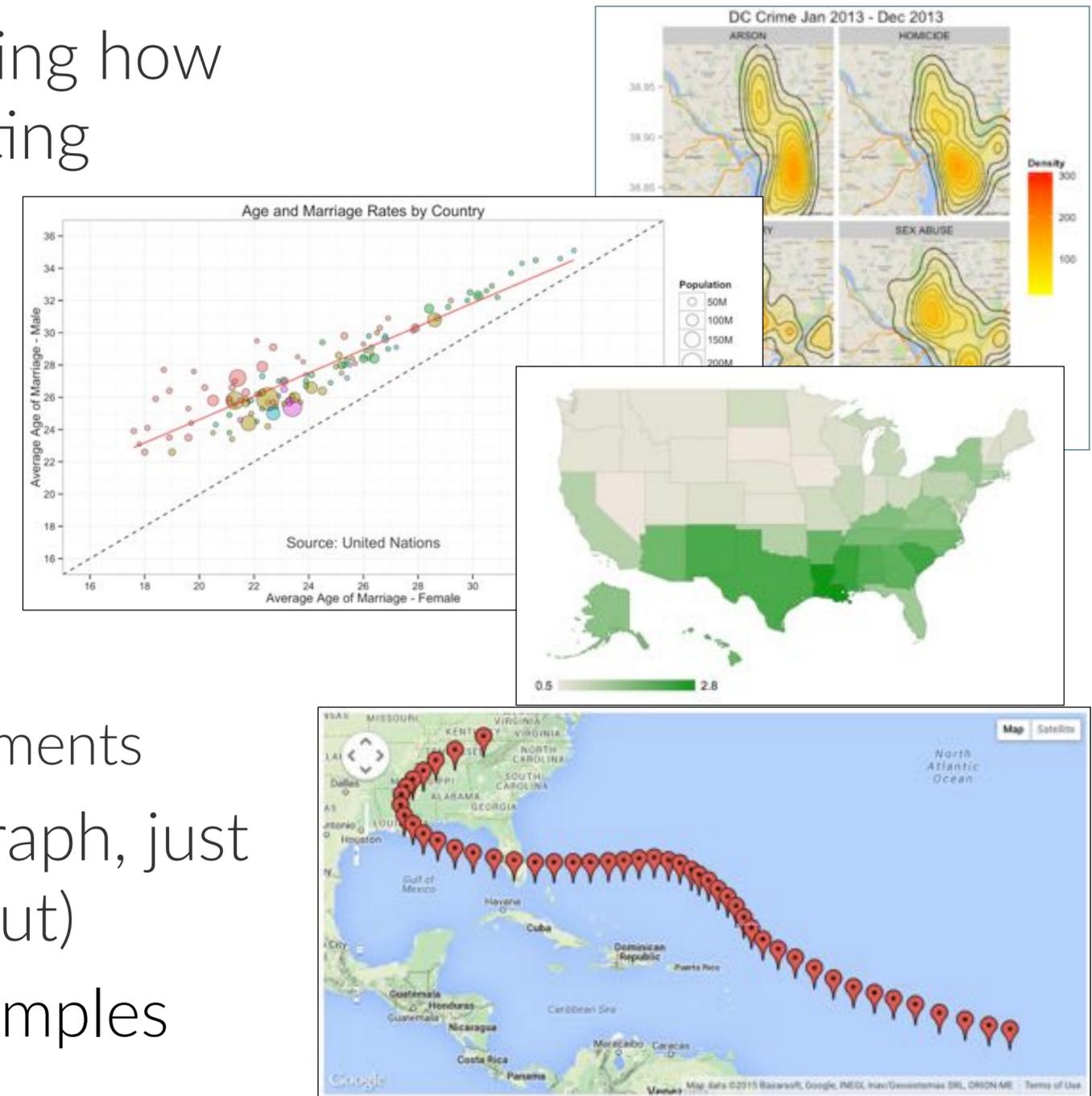
Exploratory data analysis

- R is a powerful tool for EDA because the graphics tie in with the functions used to analyze data
- You can create graphs without breaking your train of thought as you explore your data
- Visualization is an iterative process
 1. Analyze
 2. Manipulate
 3. Graph
 4. Repeat



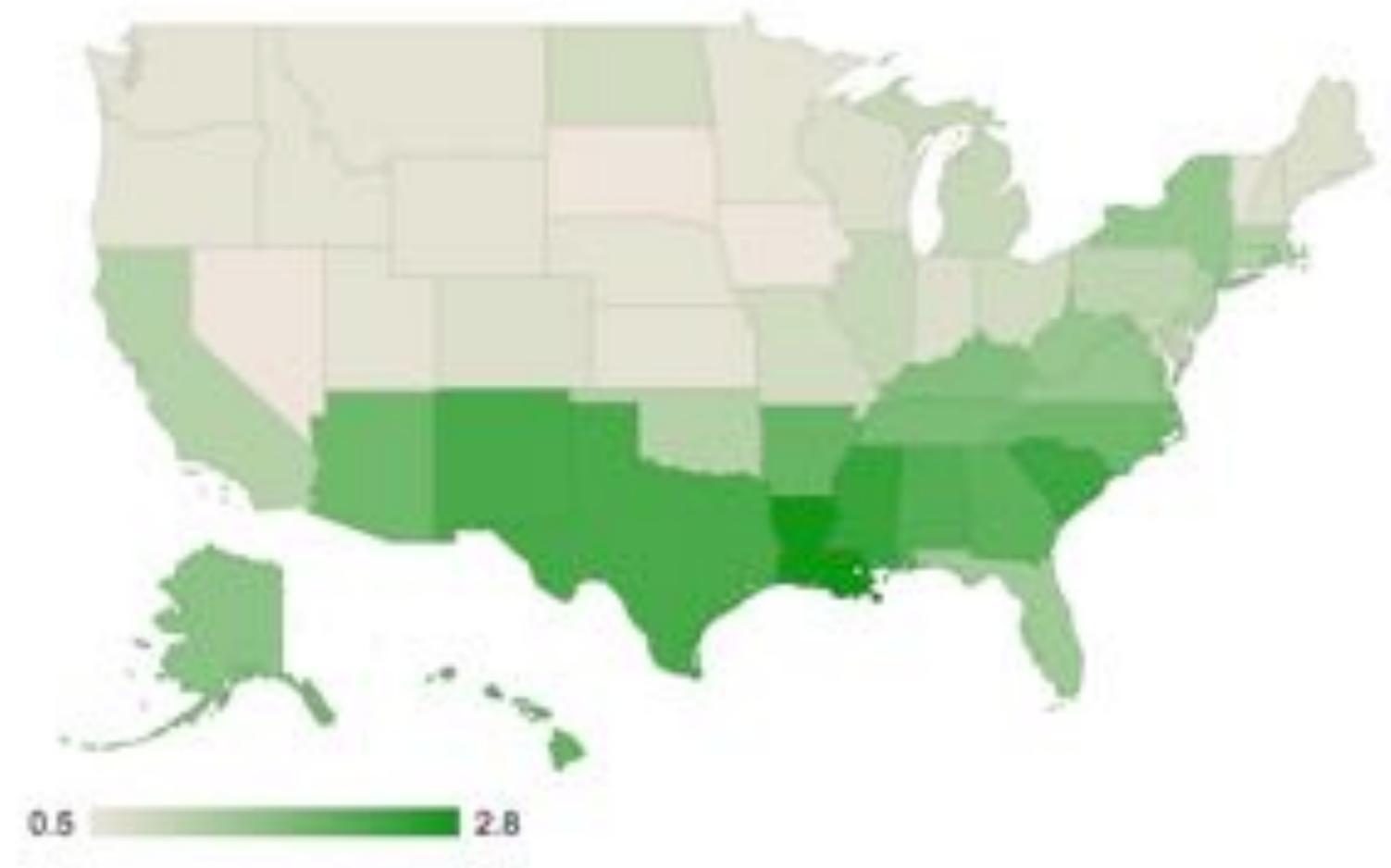
What can you ultimately do

- The key to visualization in R is understanding how your data maps to the image you are creating
 - How R interprets your data
- Visualization types
 - Maps
 - Dynamic visualizations
 - Web based & interactive
 - 3D
 - Composite graphs made of many different elements
- Repeatable (no need to ever re-create a graph, just re-use the commands you already typed out)
- Use ??visualization to see other examples



Visualization in R

- R comes with basic plotting functionality
- More advanced visualizations are done through packages
 - `ggplot2`
 - `ggpairs`
 - `ggmaps`
 - `rgl`
 - `googleVis`
 - `lattice`



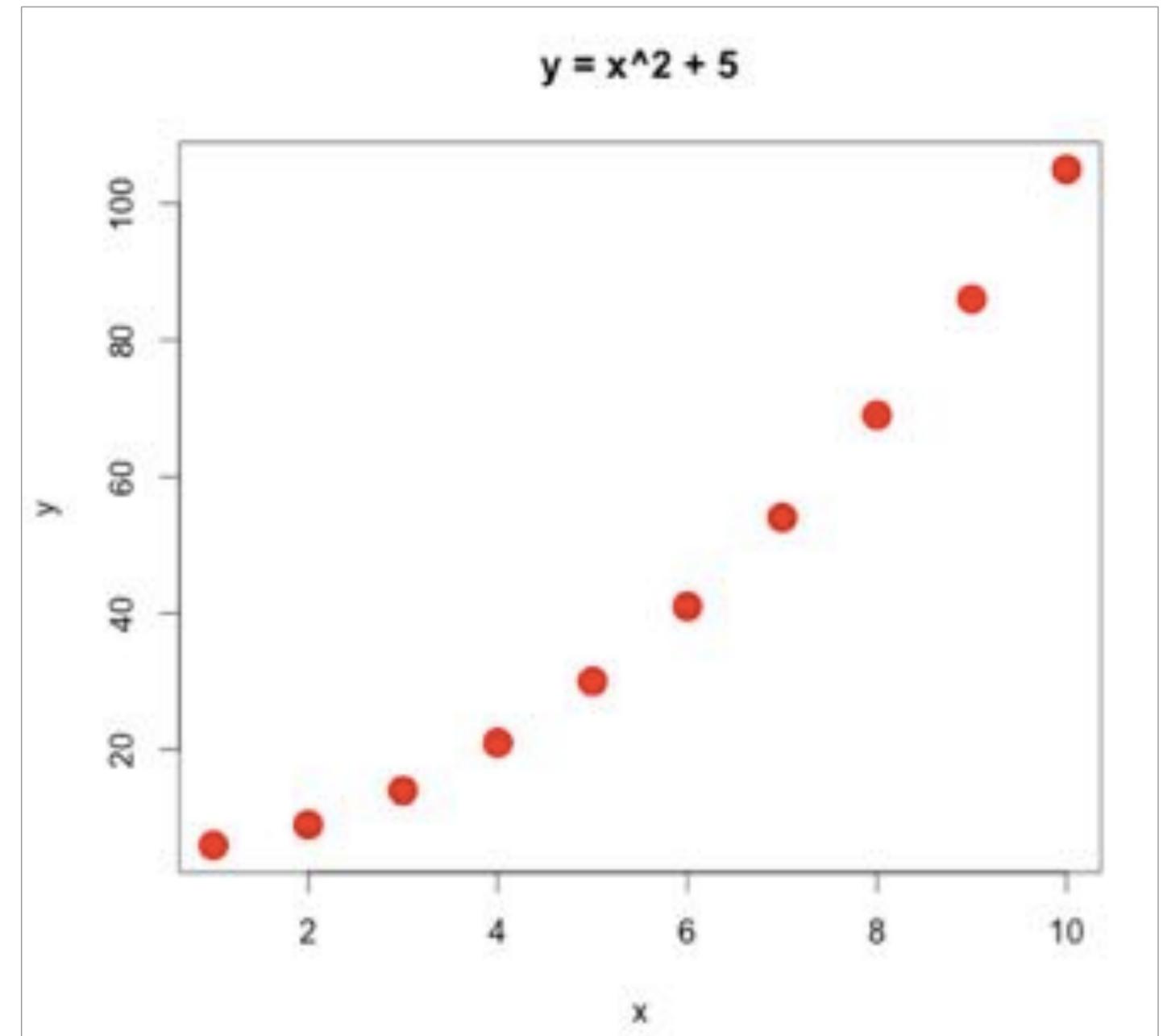
Basic scatter plot

- x , y – coordinates
- `xlab`, `ylab`, `main` – graph labels
- `lwd` – size of dot / line
- `col` – colors, "red", "blue", "green",...

```
# Suppose we want to plot  $y = x^2 + 5$ 
x = seq(1:10)
y = x^2 + 5

plot(x, y,
     lwd = 10,
     col = "red",
     main = "y = x^2 + 5")
```

Script

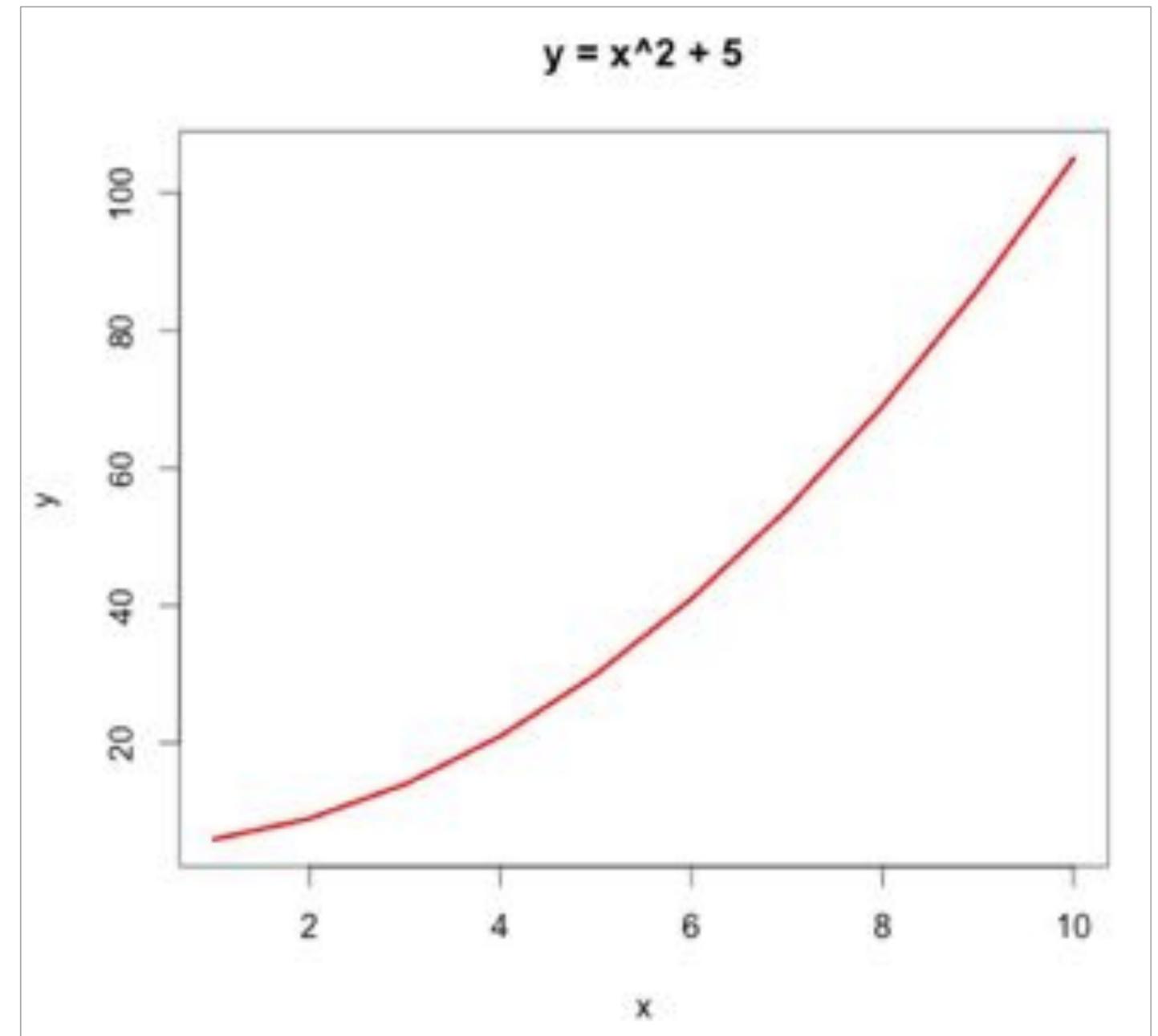


Basic line plot

- x , y – coordinates
- `xlab`, `ylab`, `main` – graph labels
- `lwd` – size of dot / line
- `type` – line, points or both
 - "l" – line
 - "b" – line and dots
- `col` – colors, "red", "blue", "green",...

```
# Suppose we want to plot  $y = x^2 + 5$ 
x = seq(1:10)
y = x^2 + 5
plot(x, y,
     xlab = "x",
     ylab = "y",
     main = "y = x^2 + 5",
     lwd = 3,
     type = "l",
     col = "red")
```

Script

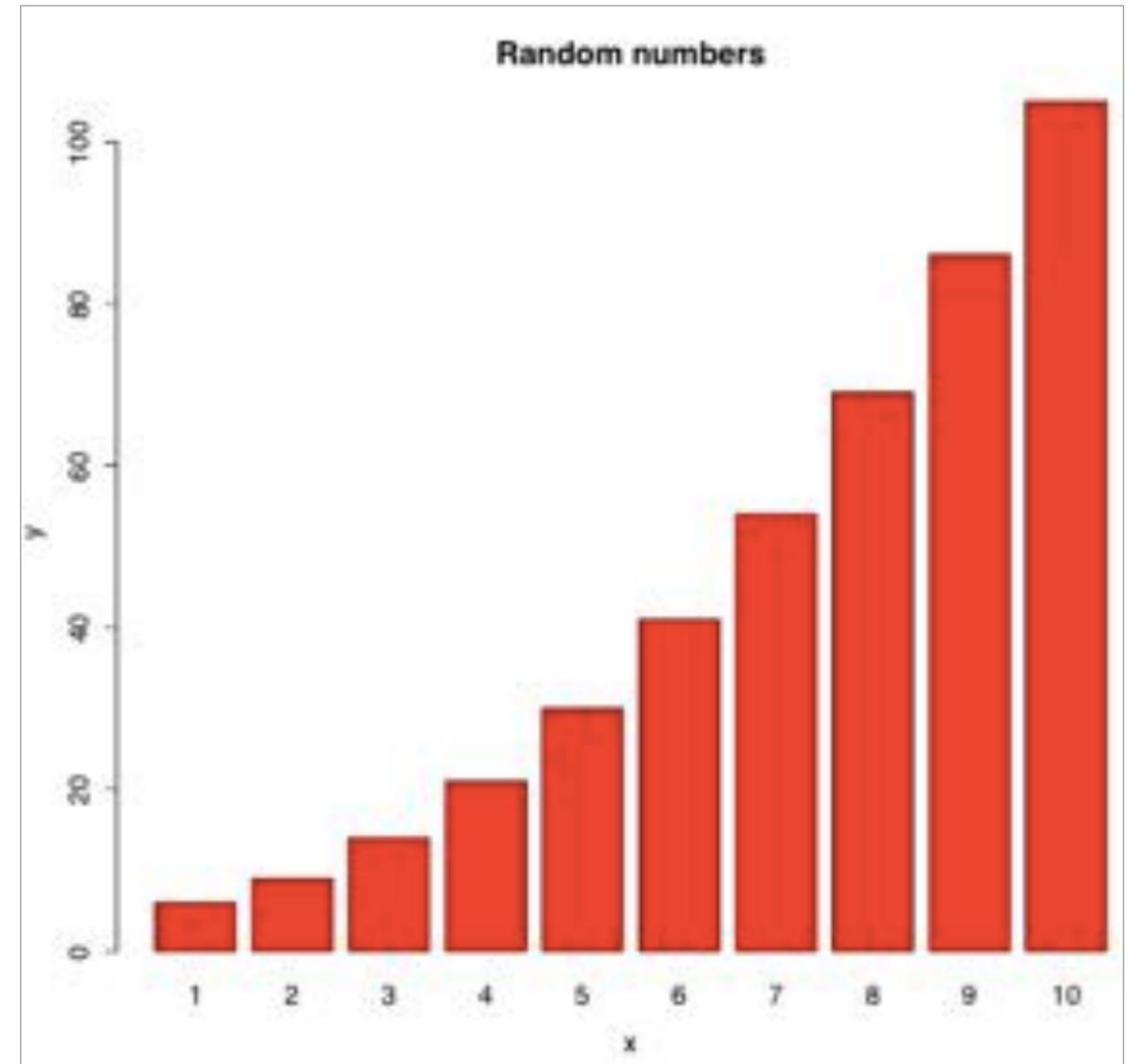


Basic bar plot

- height - value of each bar
- xlab, ylab, main - graph labels
- names.arg - labels to add to the x axis
- col - colors, "red", "blue", "green",...

```
# Suppose we want to plot 10 random
# numbers using the formula  $y = x^2 + 5$ 
x = seq(1:10)
y = x^2 + 5
barplot(height = y,
        width = 1,
        xlab = "x",
        ylab = "y",
        main = "Random numbers",
        names.arg = x,
        col = "red")
```

Script

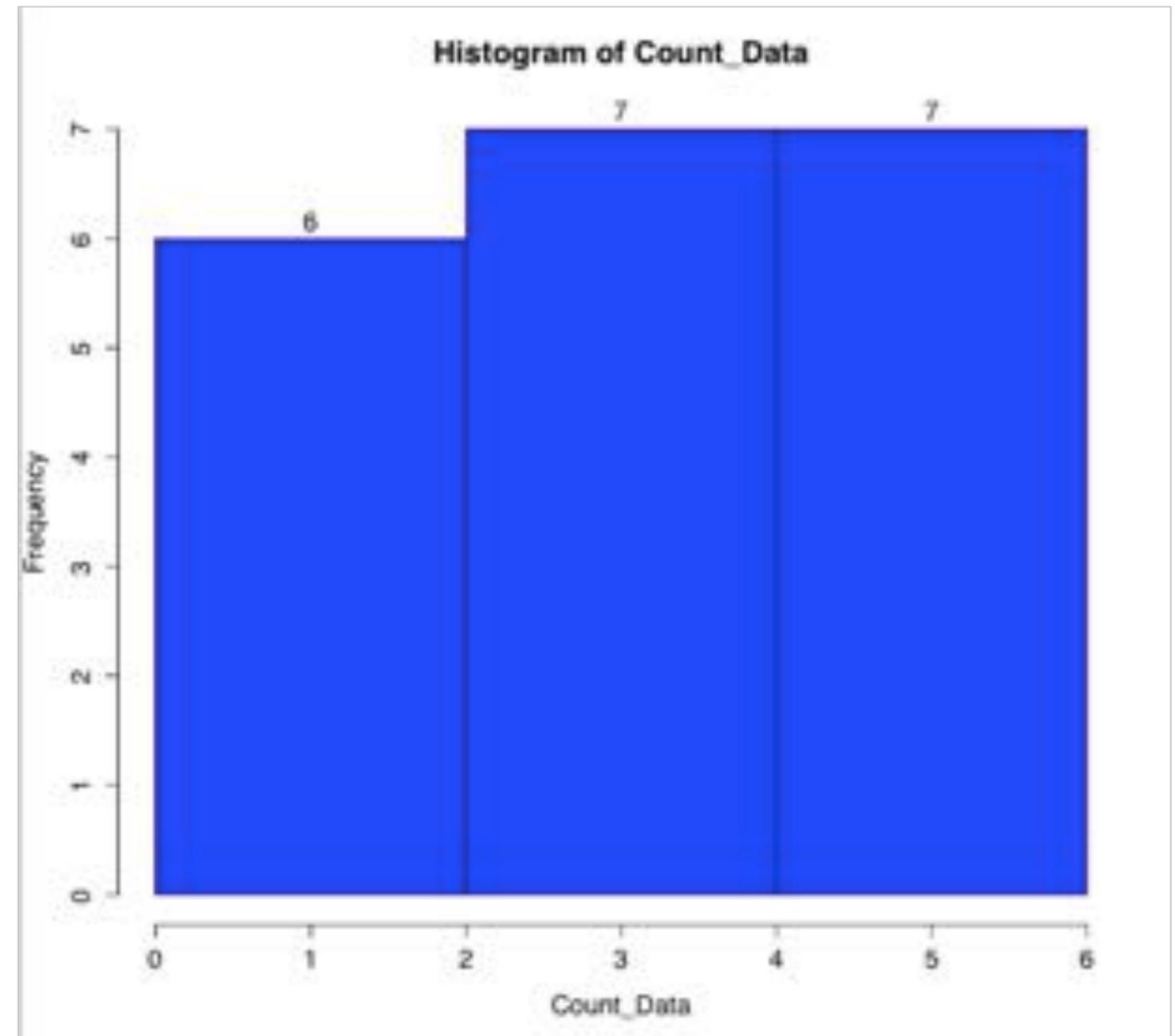


Basic histogram

- col - colors, "red", "blue", "green",...
- xlab, ylab, main - graph labels
- labels - labels on top of the columns
- breaks - allows you to specify a custom number of groupings

Script

```
# Let's plot a histogram of a sequence  
of numbers  
Count_Data = c(1, 1, 1, 1, 1,  
               2,  
               3, 3, 3, 3,  
               4, 4, 4,  
               5, 5, 5, 5, 5, 5, 5)  
  
hist(Count_Data,  
     col = "blue",  
     labels = TRUE,  
     breaks = 2)
```

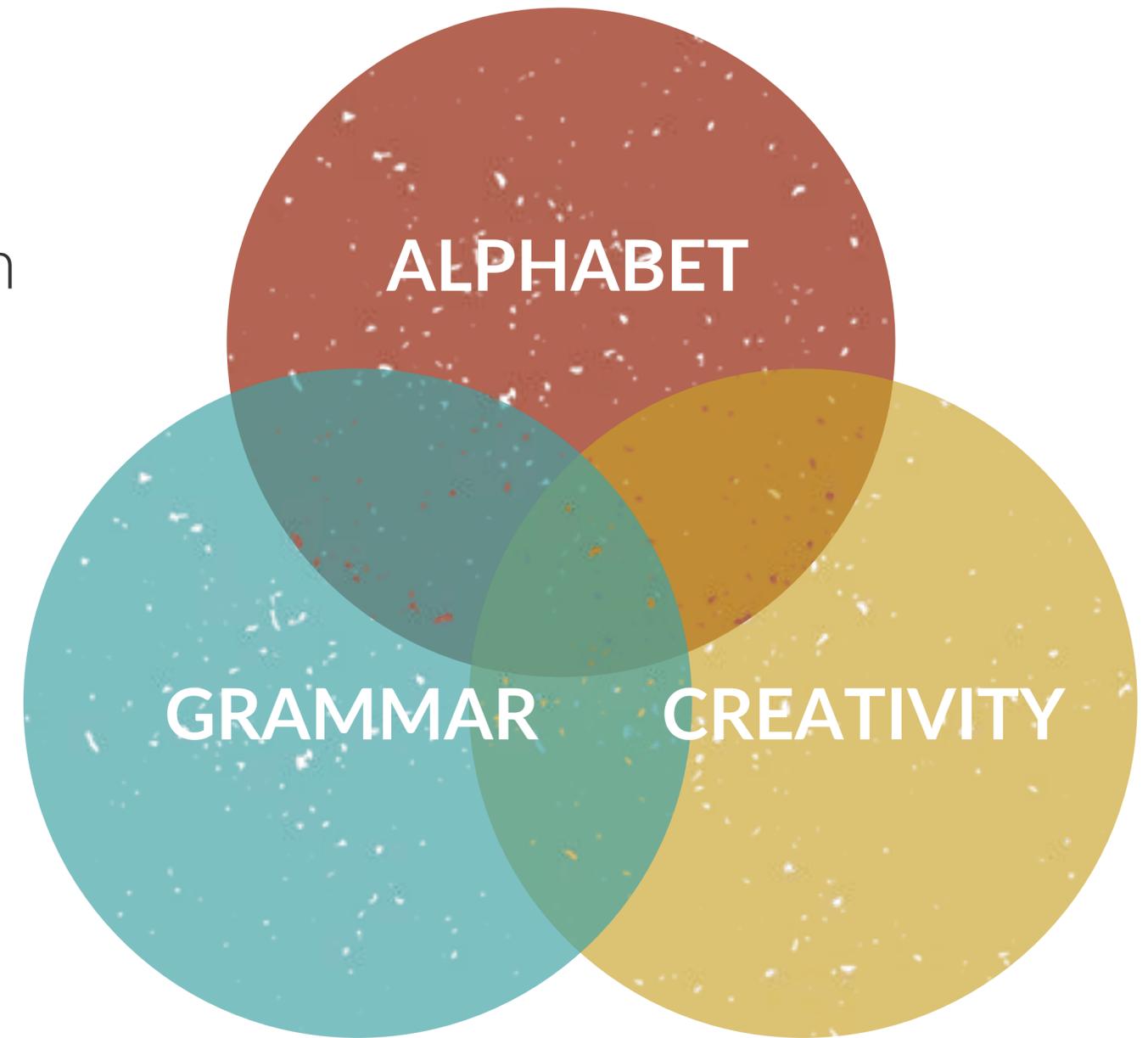
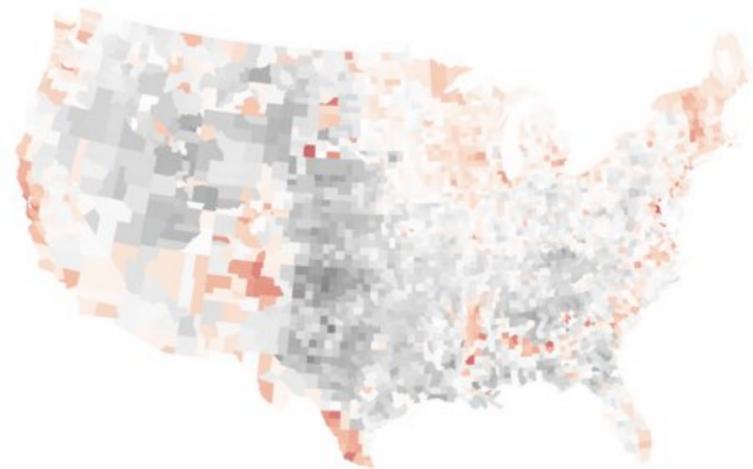


Functions for basic graphs

Chart type	Command
Line chart	<code>plot()</code>
Bar chart	<code>barplot()</code>
Histogram	<code>hist()</code>
Boxplot	<code>boxplot()</code>
Density plot	<code>density()</code>
Pie chart	<code>pie()</code>
Scatter plot	<code>plot()</code>
Heat map	<code>image()</code>

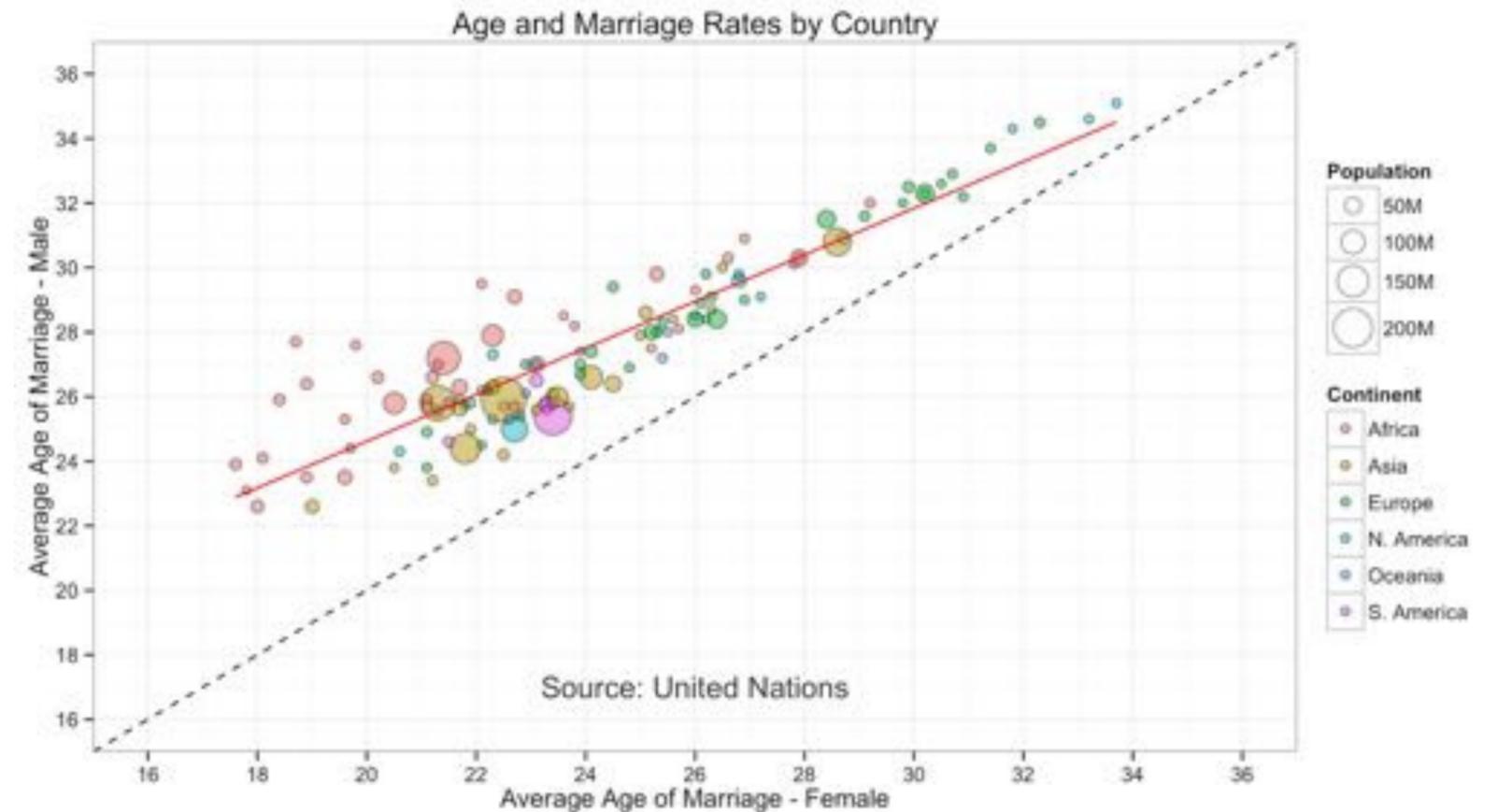
Grammar of graphics: ggp1ot2

- R comes with basic plotting functionality
- More advanced visualizations done through packages: `ggp1ot2`
- `ggp1ot2` allows the user to create flexible visualizations by combining many elements into a single image

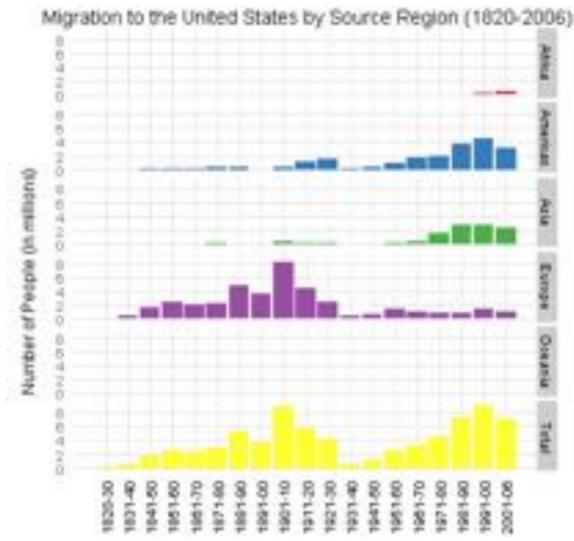


Why ggplot2

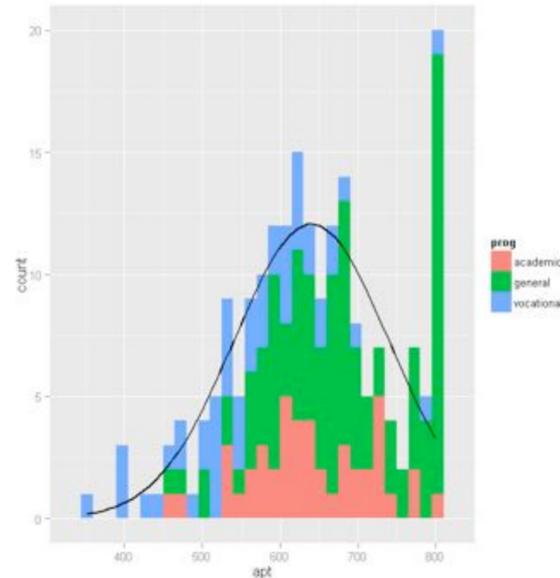
- **Explore** your data efficiently
- **Communicate** a visual story flexibly and efficiently
- **Layer** raw, summarized and contextual data
 - Demonstrate relationships
- **Reproduce** and extend your work easily



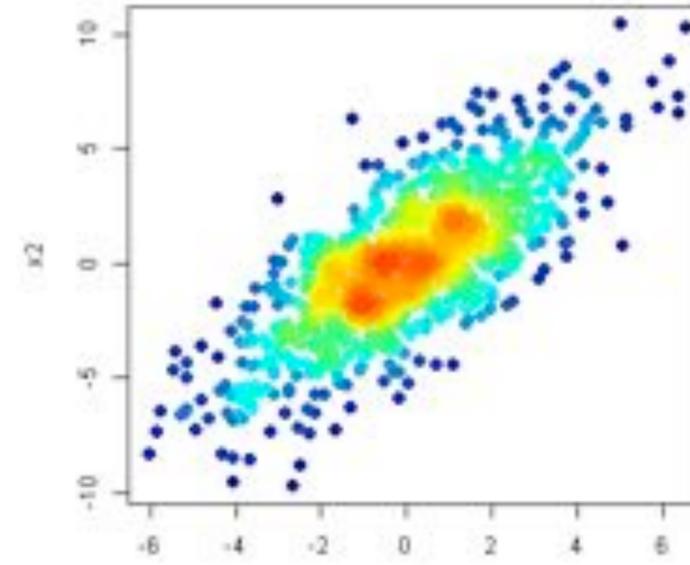
Combine basic plots for enhanced effect



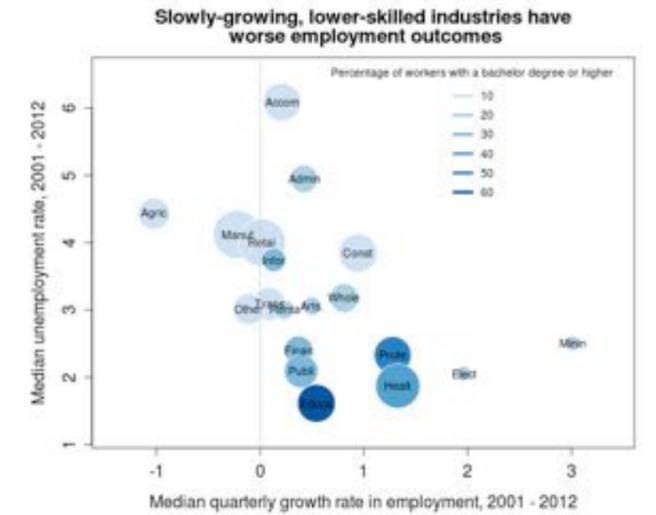
Bar plot



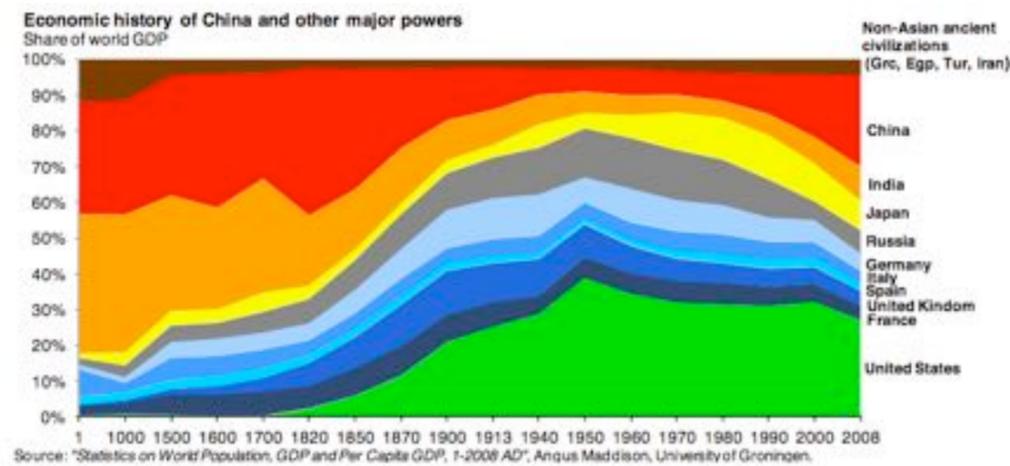
Histogram



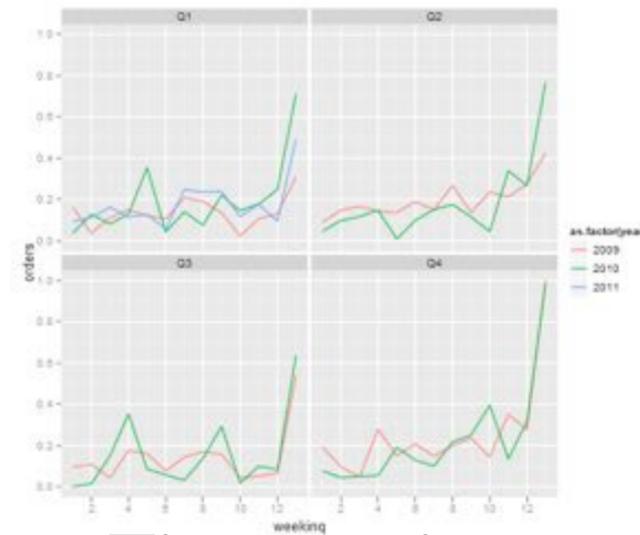
Scatterplot



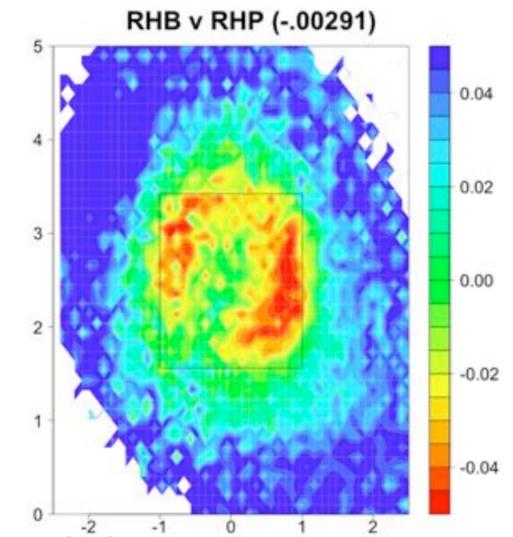
Bubble plot



Area plot

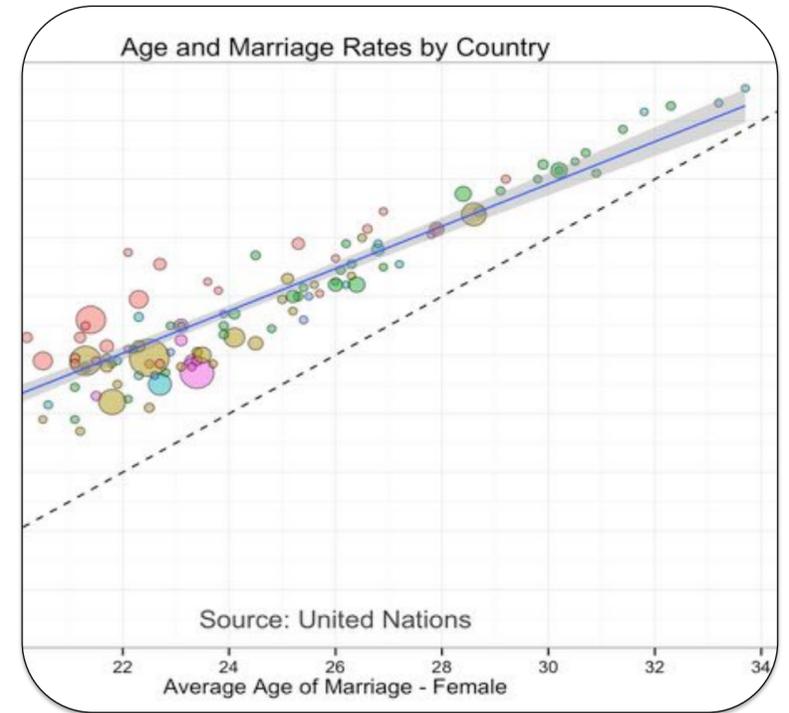
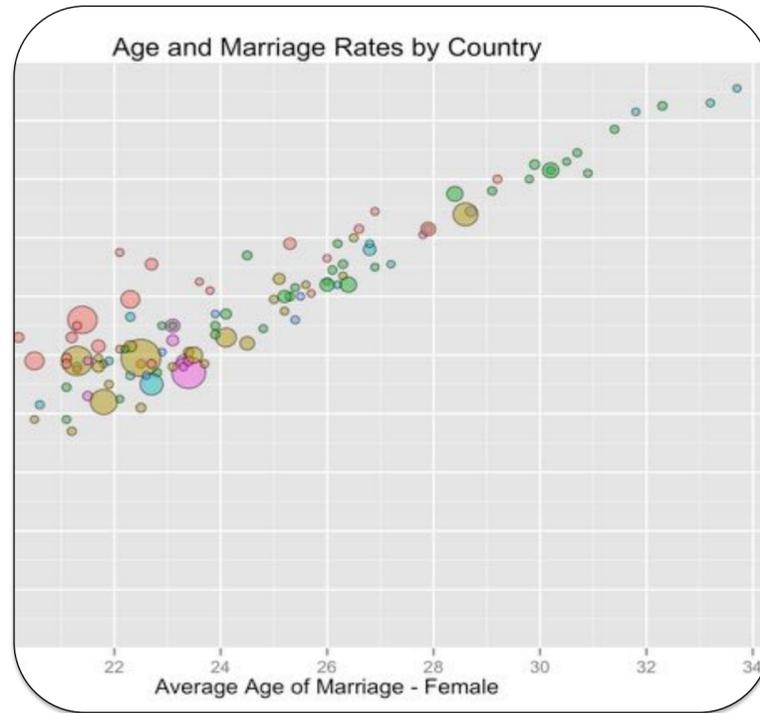
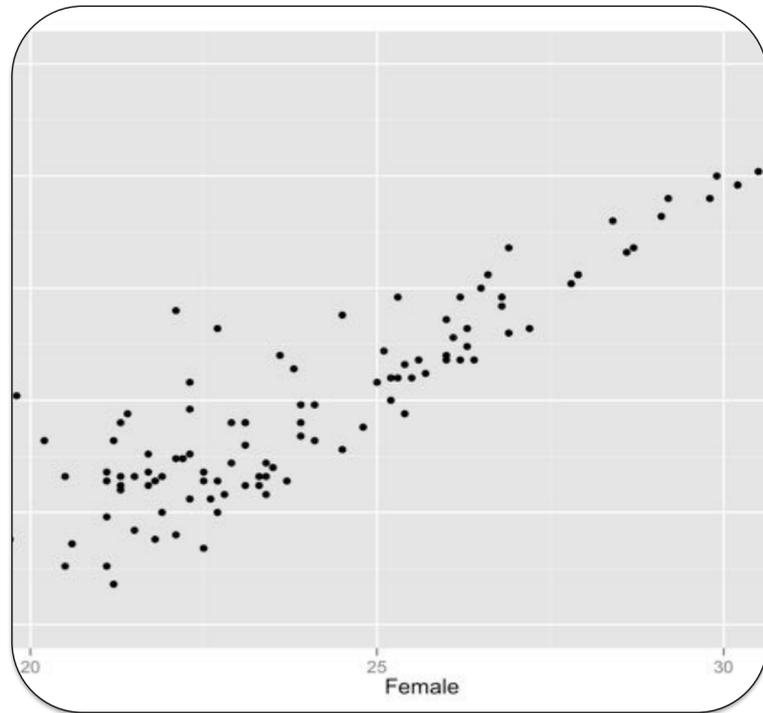


Time series



Heat map

ggplot2: building a graph



1. Specify data
2. Link data to visuals
3. Assign shapes

4. Adjust vis. effects
5. Adjust axes
6. Adjust legend

7. Customize theme
8. Layer statistics
9. Overlay text

Reading in data and loading library

Script

```
# Install ggplot2
install.packages("ggplot2")

# Loading the library ggplot2
library(ggplot2)

# Learning what ggplot2 can do
library(help = ggplot2)

# Reading in files, windows uses the C:/ directory, Mac uses the ~/ directory
setwd("path to files")
US_GDP = read.csv("US GDP.csv")

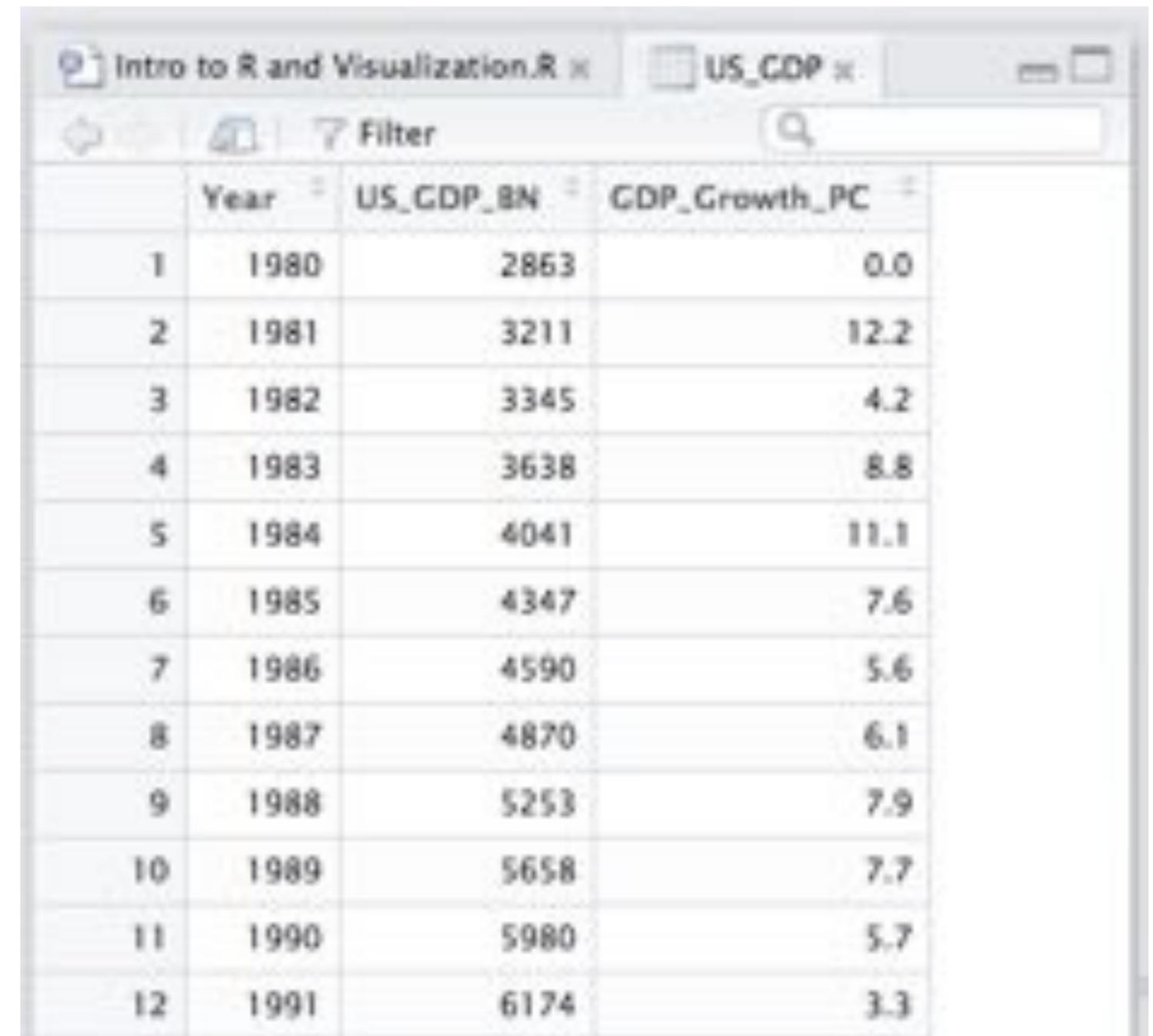
# View the file
View(US_GDP)
```

- Help: <http://docs.ggplot2.org/current/>

U.S. GDP data set

What data is available?

- U.S. **GDP** by year in \$USD billions
- U.S. GDP **growth rate** by year in percentages
- Years: 1980 - 2014
- Source: U.S. Federal Reserve



The screenshot shows an R Studio window with a data table titled 'US_GDP'. The table has four columns: 'Year', 'US_GDP_BN', and 'GDP_Growth_PC'. The data rows are numbered 1 through 12, corresponding to the years 1980 through 1991. The values for US_GDP_BN and GDP_Growth_PC are displayed in the respective columns.

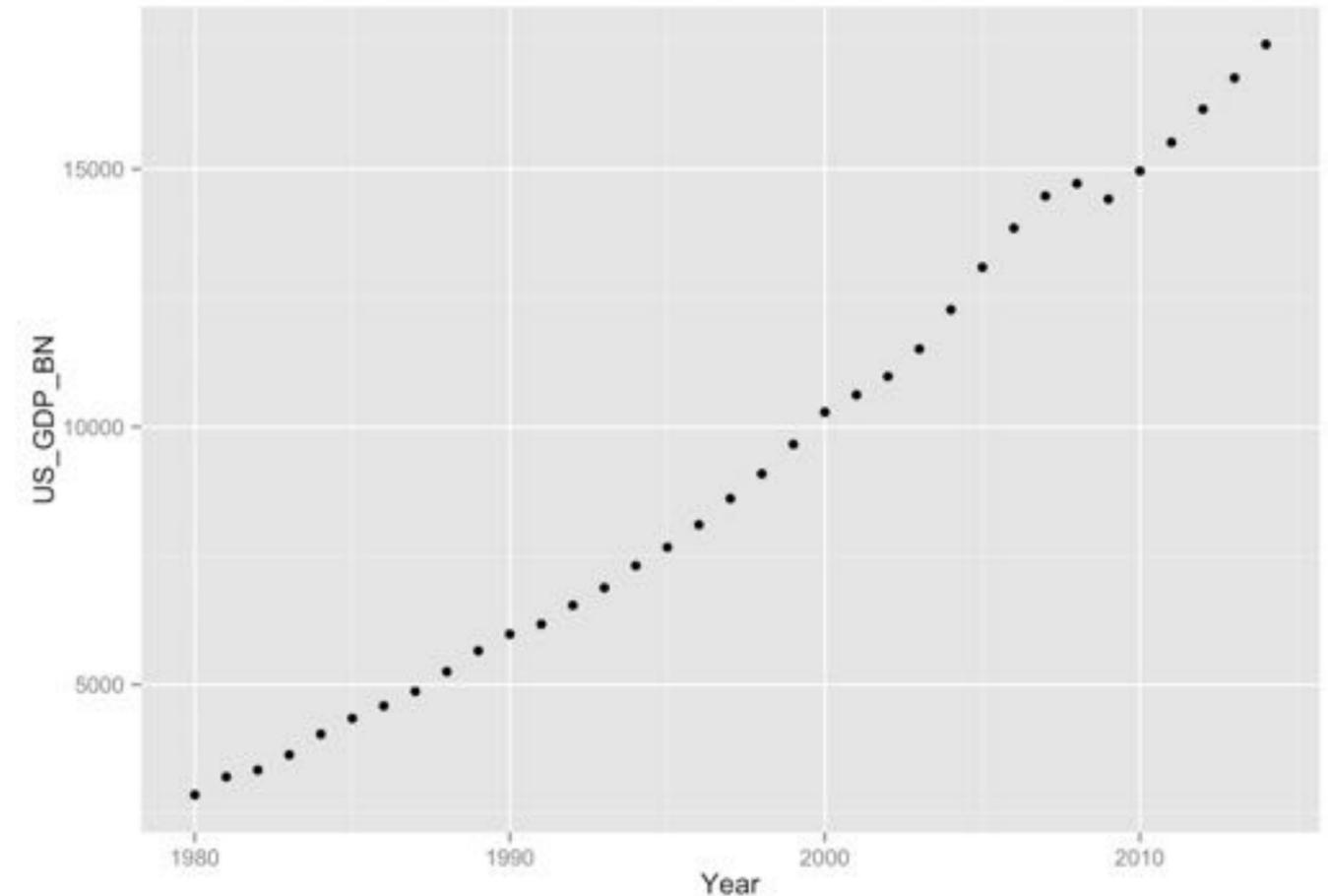
	Year	US_GDP_BN	GDP_Growth_PC
1	1980	2863	0.0
2	1981	3211	12.2
3	1982	3345	4.2
4	1983	3638	8.8
5	1984	4041	11.1
6	1985	4347	7.6
7	1986	4590	5.6
8	1987	4870	6.1
9	1988	5253	7.9
10	1989	5658	7.7
11	1990	5980	5.7
12	1991	6174	3.3

Basic scatter plot

```
ggplot(US_GDP,  
       aes(x = Year,  
           y = US_GDP_BN)) +  
  geom_point()
```

Script

- `aes` – tells R how to map / assign the data
- `geom_point()` – tells R to **use points** to display the data

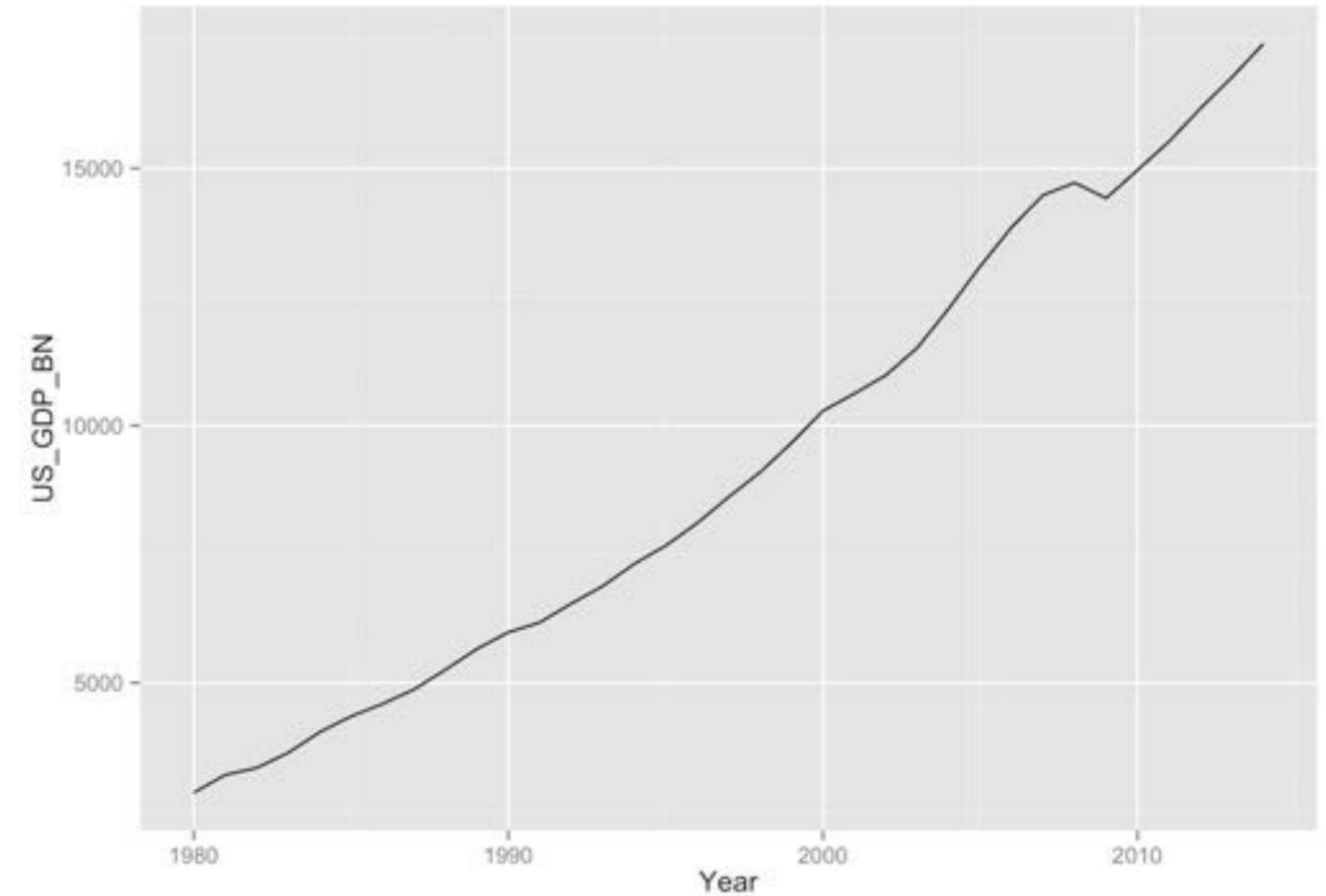


Basic line plot

```
ggplot(US_GDP,  
       aes(x = Year,  
           y = US_GDP_BN)) +  
  geom_line()
```

Script

- aes – tells R how to map / assign the data
- **geom_line()** – tells R to use a line to display the data

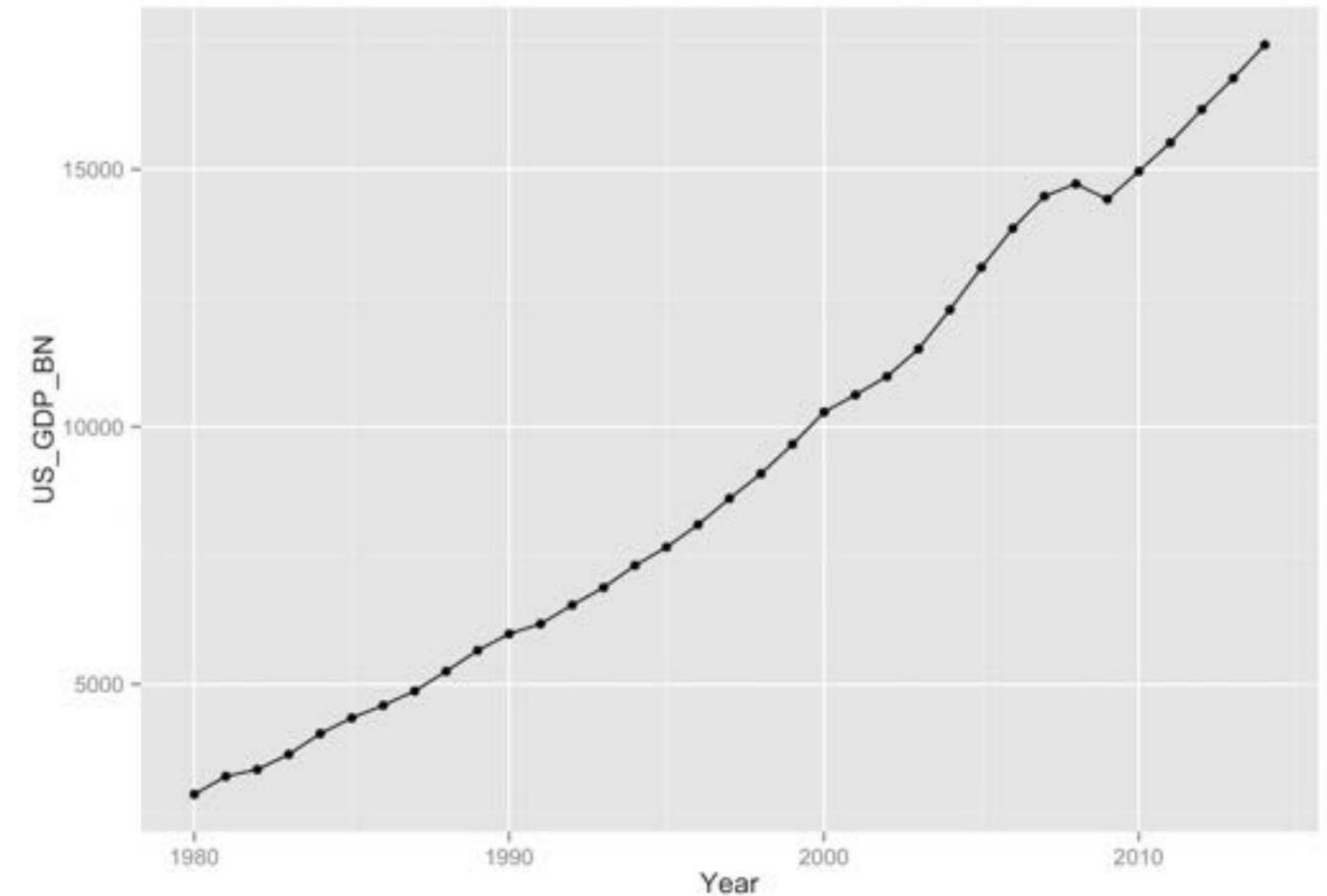


Scatter plot + line plot

```
ggplot(US_GDP,  
       aes(x = Year,  
           y = US_GDP_BN)) +  
  geom_line() +  
  geom_point()
```

Script

- **geom_line()** – tells R to use **a line** to display the data
- **geom_point()** – tells R to **use points** to display the data



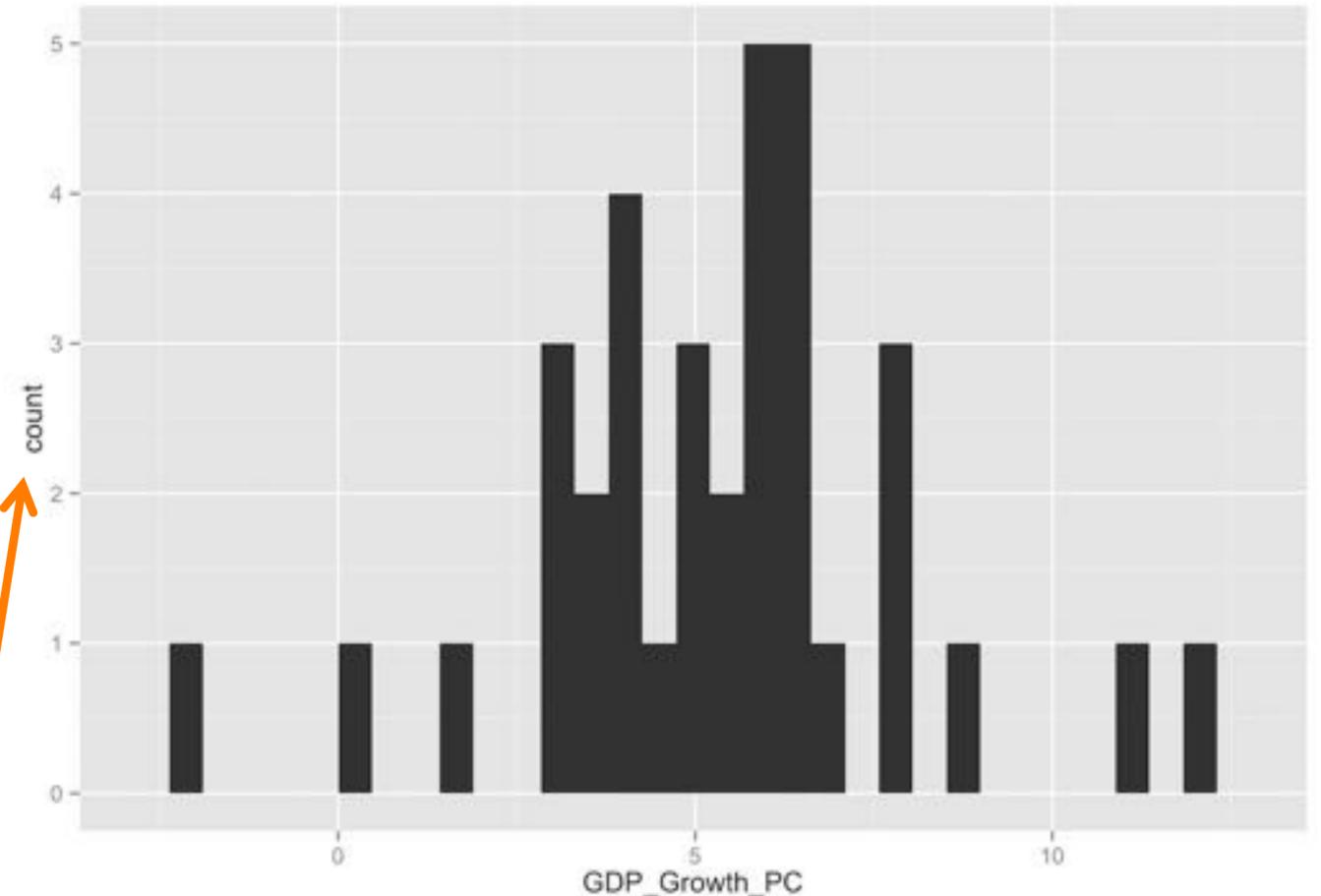
There are now 2 layers on the graph!

Basic histogram

```
ggplot(US_GDP, aes(x = GDP_Growth_PC)) +  
  geom_bar()
```

Script

- aes – note that we’re only including x-axis values
- **geom_bar()** – tells R to use **bars** to represent the data
- When given numeric data, the **default setting** of `geom_bar()` is to **bin (put data into groups)** the data and plot the number of instances each bin occurs -- a histogram



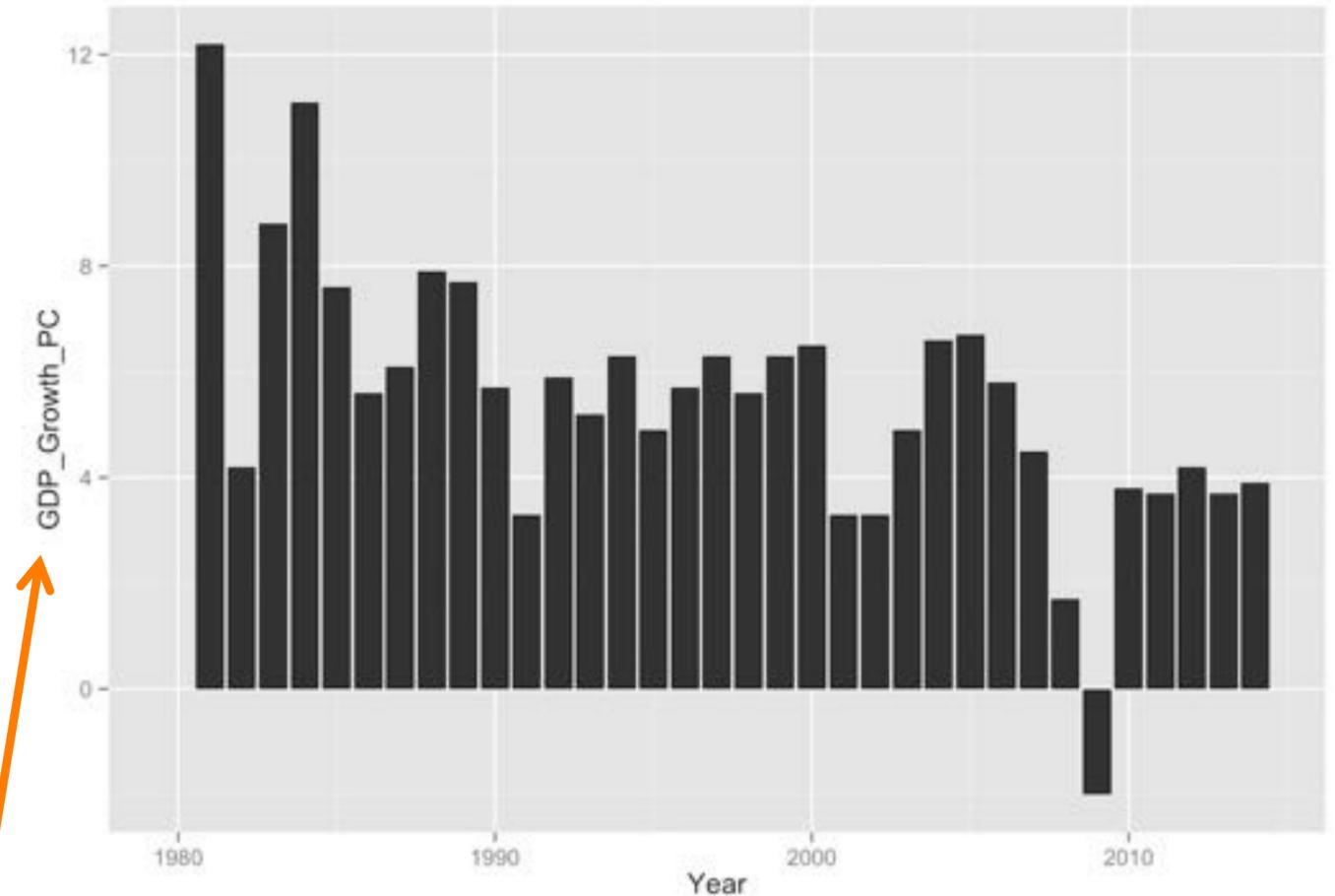
*A histogram shows the frequency of occurrence of each value on the x-axis.
This is a great way to see how often U.S. GDP grows at various rates.*

Basic bar plot

```
ggplot(US_GDP,  
  aes(x = Year,  
      y = GDP_Growth_PC)) +  
  geom_bar(stat = "identity")
```

Script

- `geom_bar()` – tells R to use **bars** to represent the data
- The default setting of `geom_bar()` plots the number of instances of a particular value on the x-axis
- To **use the data to determine bar height**, we need to tell R to use the **"identity" statistic** (`stat = "identity"`)



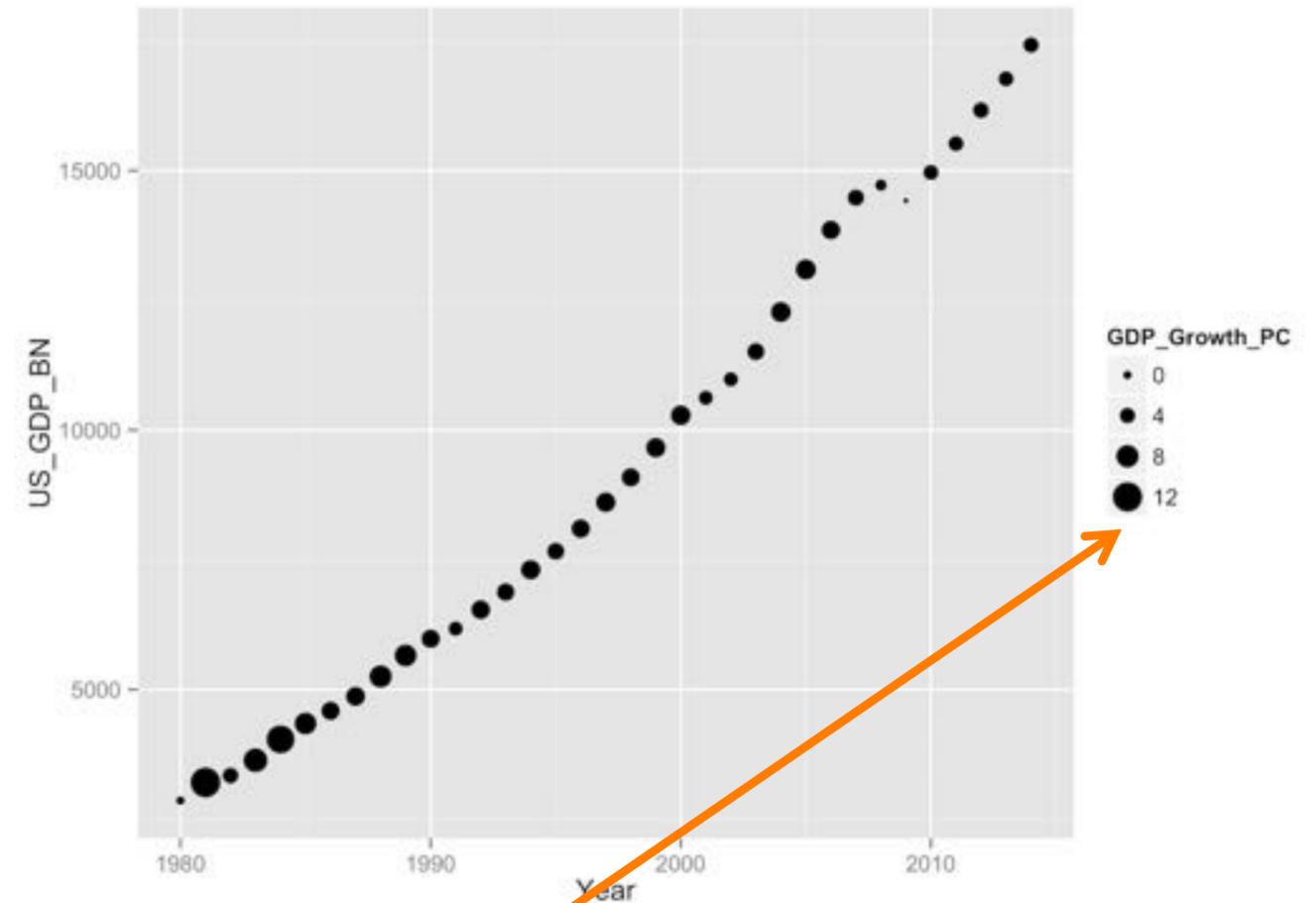
Percentage changes are usually best expressed with bars or area graphs, which are less likely to mislead your audience compared to a line graph

Bubble plot

```
ggplot(US_GDP, aes(x = Year,  
                  y = US_GDP_BN,  
                  size = GDP_Growth_PC)) +  
  geom_point()
```

Script

- The **size** argument tells R to interpret the `GDP_Growth_PC` column as a third variable *related to the size of the data markers*
- **geom_point()** – tells R to use *points* to represent the data



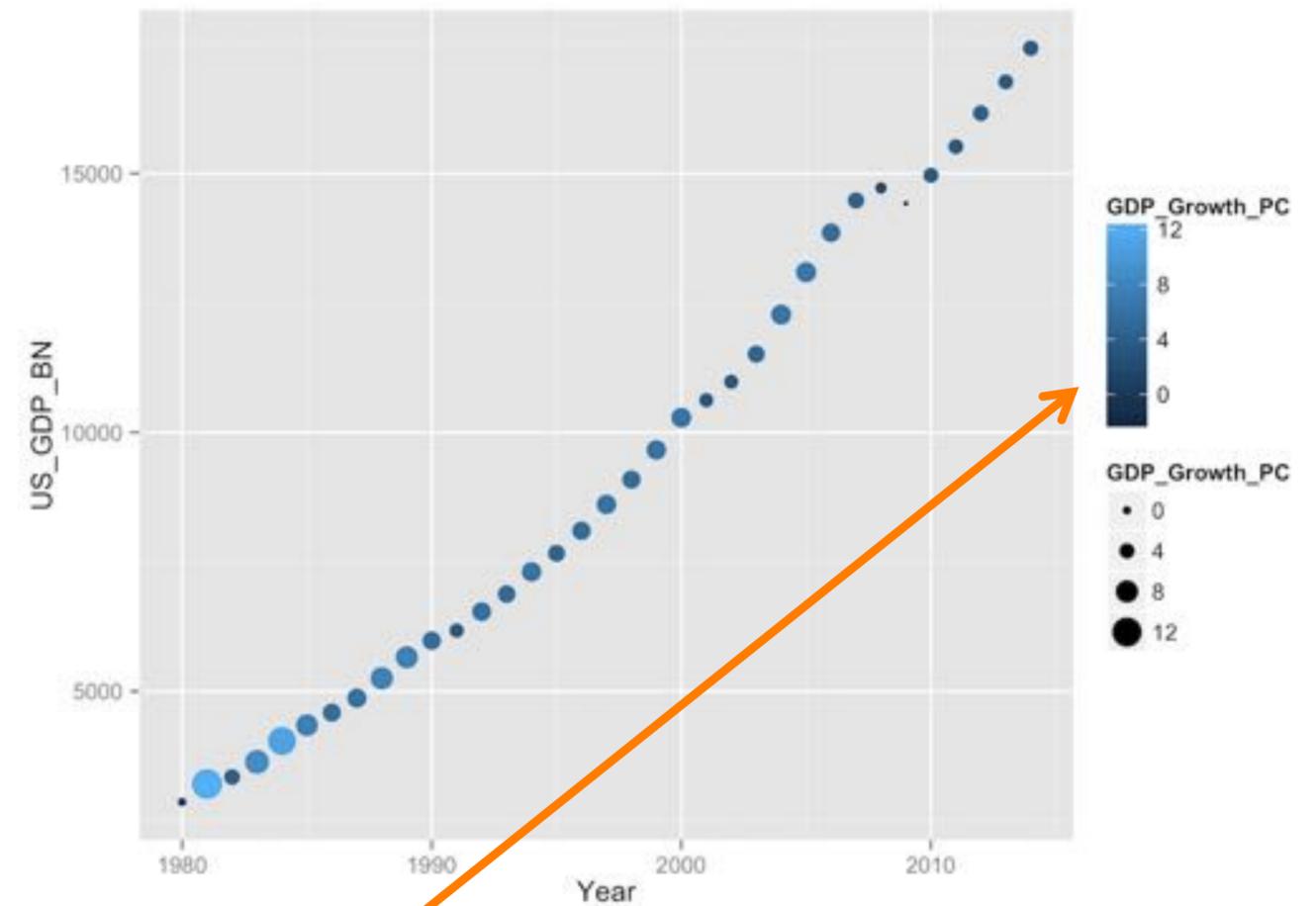
A bubble plot allows you to plot a 3rd variable on a 2D graph

Colored bubble plot

```
ggplot(US_GDP, aes(x = Year,  
                  y = US_GDP_BN,  
                  size = GDP_Growth_PC,  
                  color = GDP_Growth_PC)) +  
  geom_point()
```

Script

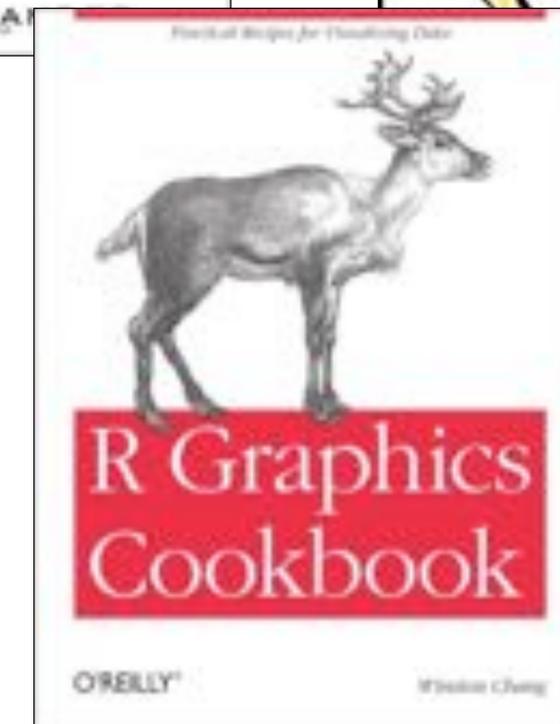
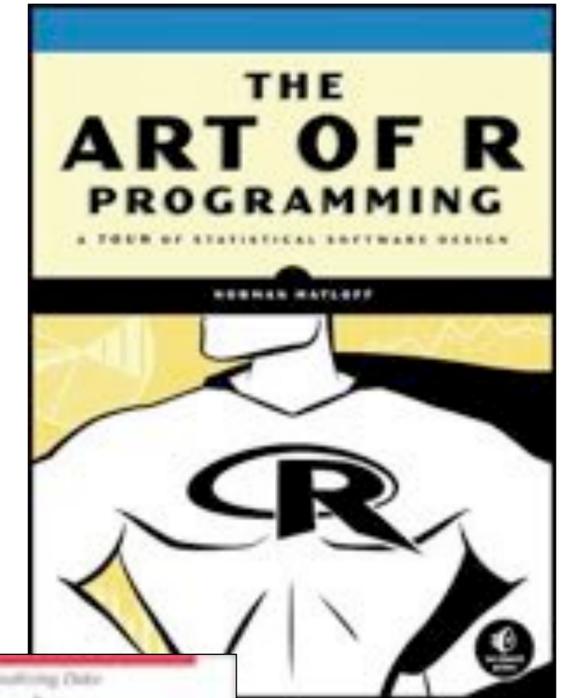
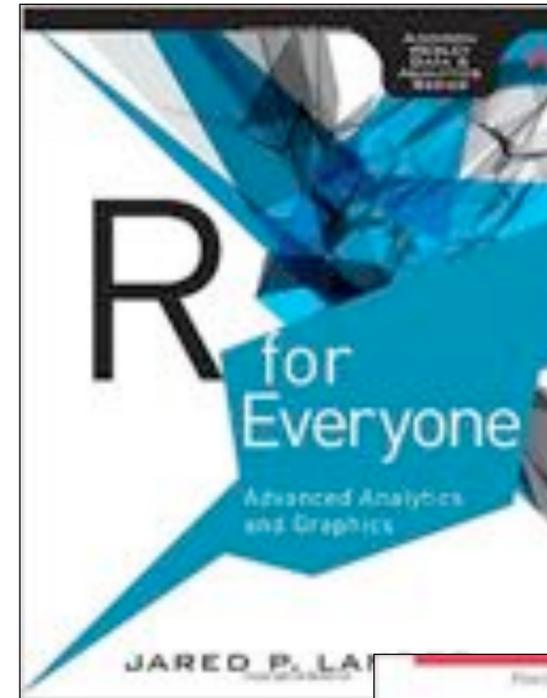
- The **color** argument tells R to interpret the GDP_Growth_PC column as a third variable *distinguished by color*
- `geom_point()` – tells R to use **points** to represent the data



The prior graph was harder to read because the points were so close in size. Adjusting the color makes the chart more legible.

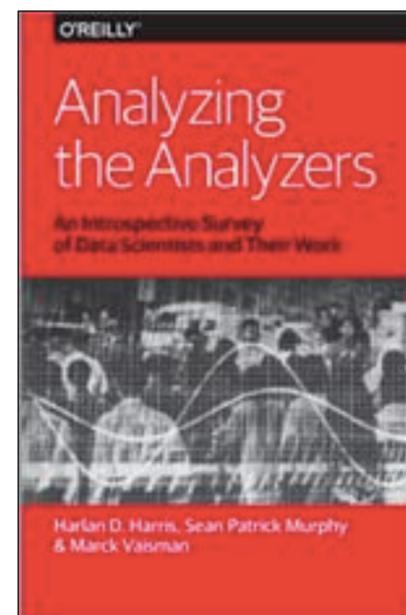
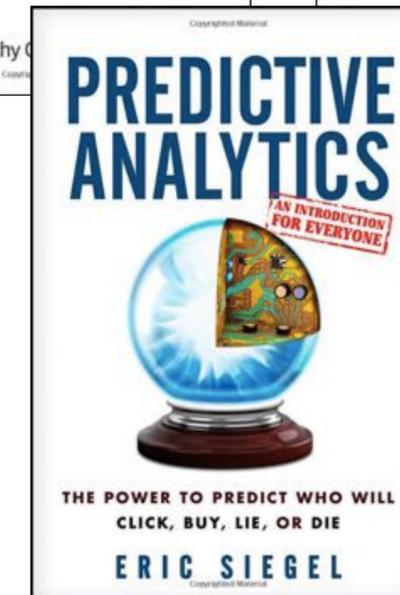
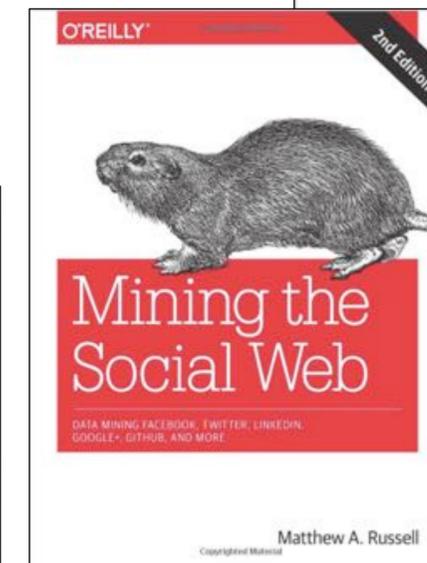
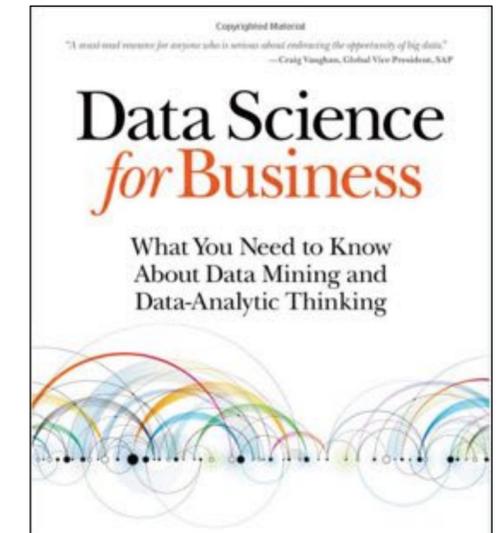
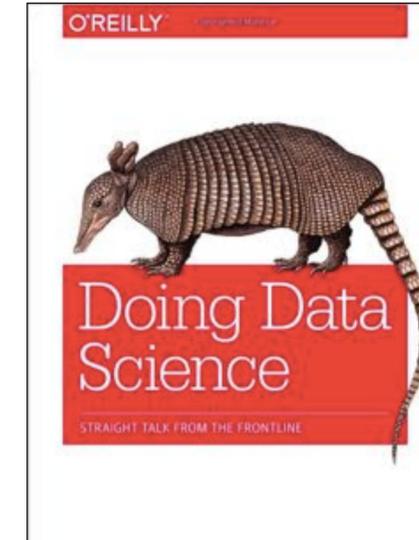
Additional R resources

1. <http://www.statmethods.net>
2. <http://ggplot2.org>
3. <https://plot.ly/r/>
4. www.r-bloggers.com
5. cran.r-project.org/
6. *The Art of R Programming* by Norman Matloff
7. *R for Everyone* by Jared P. Lander
8. *R Graphics Cookbook* by Winston Chang



Additional data science resources

1. *Doing Data Science* by Cathy O’Neil & Rachel Schutt
2. *Data Science for Business* by Foster Provost & Tom Fawcett
3. *Data Smart* by John W. Foreman
4. *Mining the Social Web* by Matthew A. Russell
5. *Predictive Analytics* by Eric Siegel
6. *Analyzing the Analyzers* by Harlan Harris, Sean Murphy and Marck Vaisman
7. www.datasciencecentral.com
8. www.kdnuggets.com



Congratulations!!



- ✓ You have learned how to code in R
- ✓ You have learned how to think about data
- ✓ You have learned how to format data
- ✓ You have learned how to build basic visualizations

The collage displays various R Studio components and visualizations:

- Keyboard Shortcut Quick Reference:** A window listing shortcuts for navigation, editing, and execution.
- Source Editor:** Shows R code for creating a factor variable: `categories = as.factor(crime_incidents_2013$OFFENSE)` and plotting it: `plot(categories)`.
- Environment:** Shows a data frame with 35826 observations and 22 variables. The 'Values' section displays a list of crime categories: "ASSAULT W/DANGEROUS WEAPON", "BURGLARY", "HOMICIDE", "MOTOR VEHICLE THEFT", "ROBBERY", "SEX ABUSE", "THEFT F/AUTO", "THEFT F/OTHER".
- Random numbers:** A bar chart showing a distribution of random numbers from 1 to 8. The y-axis ranges from 0 to 100.
- US_GDP_BN:** A scatter plot showing US GDP in billions of dollars from 1980 to 2010. The y-axis ranges from 5000 to 15000. A legend indicates that the size of the blue dots represents GDP_Growth_PC, with sizes corresponding to 0, 4, 8, and 12.

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