# Session 1: Introduction to R

## Overview

Introductions and introduction to R

Basic operations and data types

Packages

R Markdown

**Data Frames** 

### Introductions

### My background:

- Visiting associate professor at Yale
- Associate professor of Statistics, Hampshire College
- Research Affiliate, Center for Brains, Minds and Machines at MIT

#### How I use R:

- Teaching: Statistics and Data Science courses using R
- Research: Develop/use machine learning to analyze neural data
  - NeuroDecodeR package

## Introductions

### Let's do some quick introductions

- Your name
- Your academic department affiliation
- How are you interested in using R in your teaching/research?
- What you are most interested in learning in this workshop?
- Anything else you would like to share

# Workshop plan: today

Session 1: Basic programming in R

Session 2: Statistics and plots, for loops, writing functions

Session 3: Basic statistical inference and simple linear regression

Session 4: Multiple regression, logistic regression, ANOVA

# Workshop plan: tomorrow

Session 1: Data wrangling with dplyr

**Session 2**: Data visualization with ggplot

Session 3: Joining and mapping data

**Session 4**: String manipulation and/or interactive graphics

Even if you don't like today, come back tomorrow b/c that's when we're going to be doing the really fun stuff!



# Why use R?



## Data analysis software vs. programming languages

Advantages of using data analysis software (STATA, SAS, SPSS, Minitab, etc.):

• Can be easier/quicker to do the basics: you don't need to know to how to program!

Advantages of using data analysis programming languages

- Reproducibility! Can see exactly which steps were taken in an analysis.
  - Can cut down on mistakes
- Ability to extend beyond functions built into software
  - Particularly useful for computational methods that work in more situations



# Data analysis programming languages



#### **MATLAB**

- Disadvantage: Expensive
  - Makes it harder to distribute results because others might not own the software



#### Julia

Disadvantage: Newer so less of a community, less functionality implemented



#### Python

- Advantages: general purpose programming language so useful if one already knows Python of if one wants to integrate into larger software ecosystem
- Disadvantage: Not quite as good as R for data analysis



#### R

- Advantages: Rich ecosystem designed specifically for reproducible data analysis
  - Free RStudio IDE, reproducibility, powerful packages, etc.
- Disadvantages: Not good for larger software development

#### AS SEEN BY USERS OF ...







































































# What are the strengths of R?

Statistical analyses: basic and advanced

Reproducibility: R Markdown



Tidyverse: Powerful packages for graphics, data manipulations, etc.

Interactive graphics: plotly and Shiny

Demos!



## R and R Studio

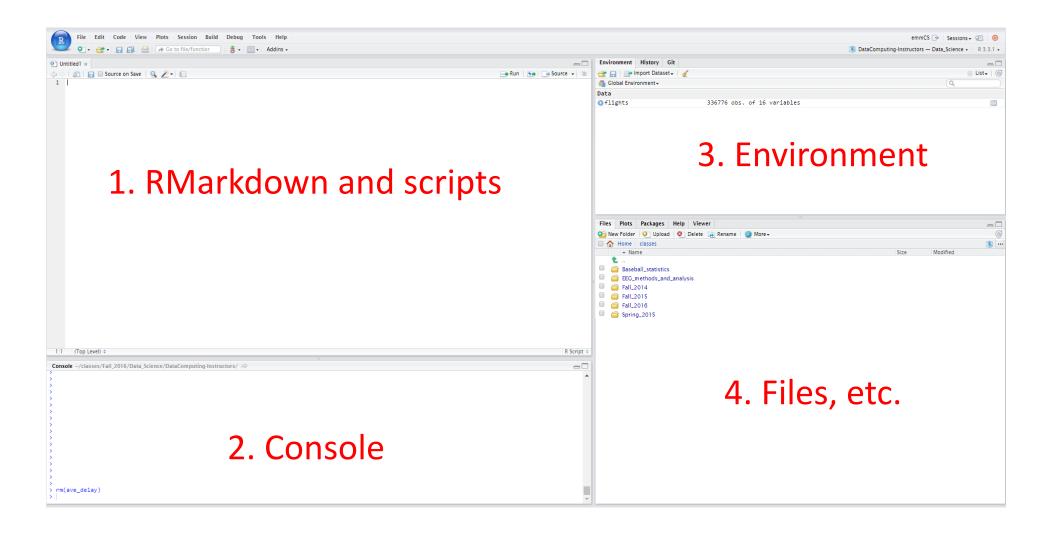
R: Engine



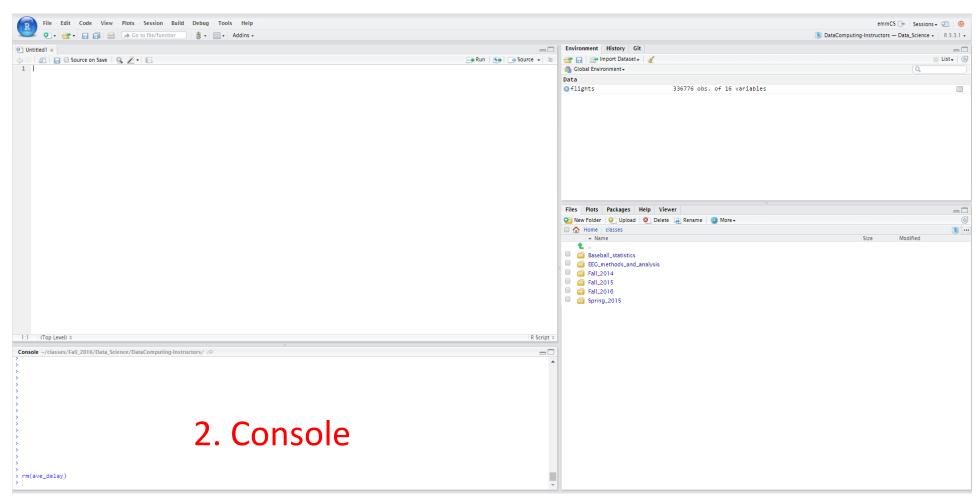
#### **RStudio: Dashboard**



# RStudio layout



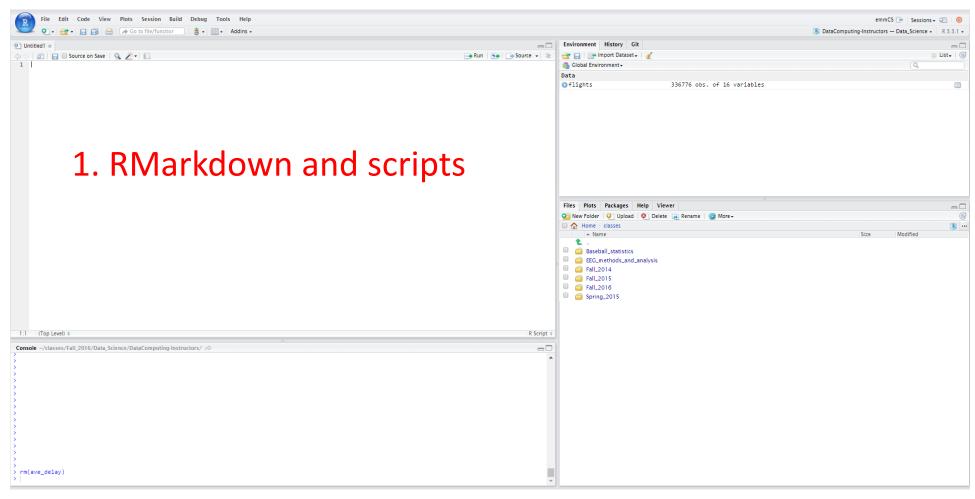
# RStudio layout



### R as a calculator

- > 2 + 2
- > 7 \* 5

## RStudio layout



#### Create a new script

File -> New File -> R Script

Save the script with a reasonable name, e.g., my\_notes.R

## R Basics

#### Arithmetic:

2 + 27 \* 5

### Assignment of values to *objects*:

> a <- 4</li>
> b <- 7</li>
> z <- a + b</li>
> z
[1] 11

Number journey...

# Number journey

```
> a <- 7
> b <- 52
> d <- a * b
> d
[1] 364
```

# Character strings and Booleans

```
> a <- 7
> s <- "s is a terrible name for an object"
> b <- TRUE
> class(a)
[1] numeric
> class(s)
[1] character
```

### **Functions**

Functions use parenthesis: functionName(x)

```
> sqrt(49)
> tolower("DATA is AWESOME!")
```

To get help

> ? sqrt

One can add comments to your code

> sqrt(49) # this takes the square root of 49

### Vectors

Vectors are ordered sequences of numbers or letters The c() function is used to create vectors

```
> v <- c(5, 232, 5, 543)
> s <- c("statistics", "data", "science", "fun")
```

One can access elements of a vector using square brackets [] > s[4] # what will the answer be?

We can get multiple elements from a vector too > s[c(1, 2)]

## Vectors continued

One can assign a sequence of numbers to a vector

- > z <- 2:10
- > z[3]

One can test which elements are greater than a value

Can add names to vector elements

```
> names(v) <- c("first", "second", "third", "fourth")
```

## Vectors continued

One can also apply functions to vectors

- > z <- 2:10
- > sqrt(z)
- > mean(z)

# Questions?



# R packages



Packages add additional functionality to R

We will use many additional packages in this class

• gplyr, ggplot2, tidyr, etc.

# R workshop package

I have written a package to allow you to more easily download material for this workshop

```
> install.packages("devtools")
```

- > devtools::install\_github("emeyers/rworkshop")
- > devtools:::initial setup()

Once you have installed this package, please run the following command to download an R Markdown document that has code for this morning session

> rworkshop::download\_class\_code(1)

### R Markdown

R Markdown (.Rmd files) allow you to embed written descriptions, R code and the output of that code into a nice looking document



Creates a way to do reproducible research!



## R Markdown

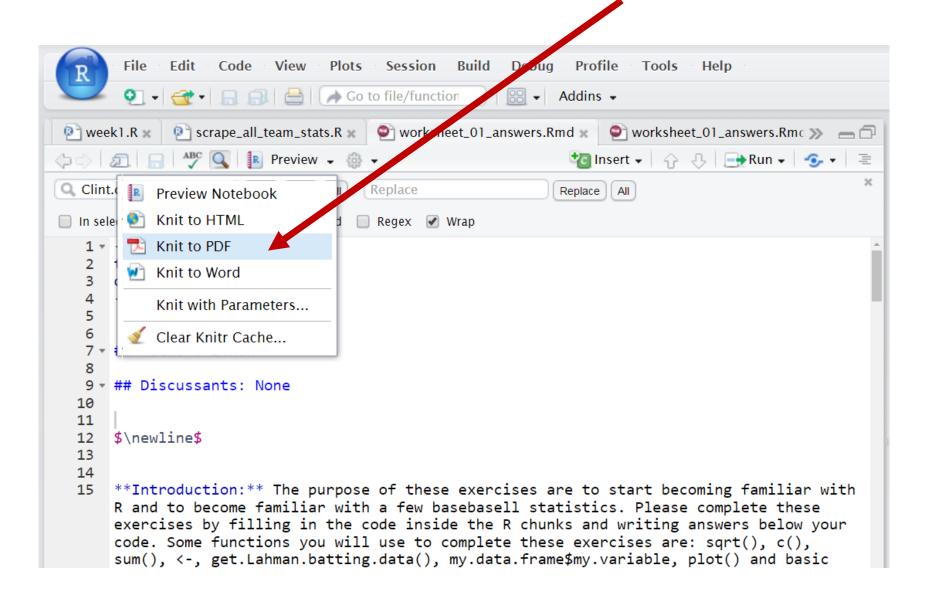
Everything in R chunks is executed as code:

```
'``{r}
  # this is a comment
  # the following code will be executed
  2 + 3
'``
```

Everything outside R chunks appears as text

# Knitting to a pdf

# Turn in a pdf or html document with your solutions to Canvas



## R Markdown

Note: When you knit, RMarkdown files <u>do not have access to</u> <u>variables in the global environment</u>, but instead have their own environment.

Why is this a good thing???

# Formatting in R Markdown

We can add formatting to text outside the code chunks

### Examples:

```
## Level 2 header
**bold**
![](https://statistics.yale.edu/sites/default/files/logo2.png)
```

### LaTeX in R Markdown

We can also add LaTeX symbols to documents using \$\symbol\$ syntax

### For example, try these:

```
$\theta$
$\hat{p}$
$\hat{\theta}$
```

Knit early and knit often to avoid errors!!!

# Avoid hard to debug code!

Only change a few lines at a time and then knit your document to make sure everything is working!

If you document isn't knitting:

- For code chunks: use the # symbol to comment out code until you can find the line of code that is giving the error message
- Outside of code chunk: cut out part of the document until it knits and then paste it back

# Questions?

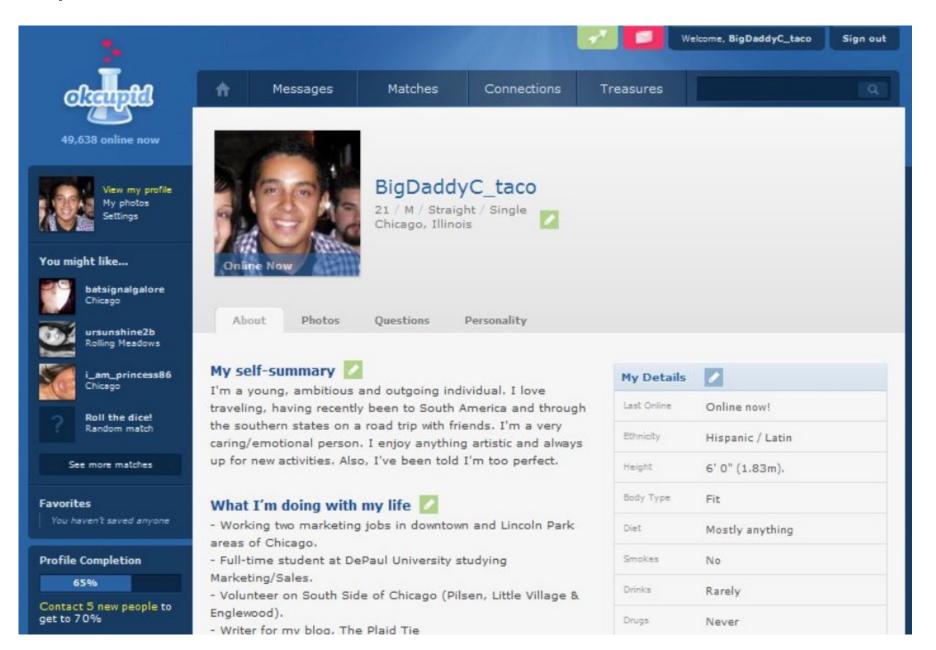


## Data frames

Data frames contain structured data

_	age <sup>‡</sup>	body_type <sup>‡</sup>	diet	drinks <sup>‡</sup>	drugs <sup>‡</sup>	education
1	22	a little extra	strictly anything	socially	never	working on college/university
2	35	average	mostly other	often	sometimes	working on space camp
3	38	thin	anything	socially	NA	graduated from masters program
4	23	thin	vegetarian	socially	NA	working on college/university
5	29	athletic	NA	socially	never	graduated from college/university
6	29	average	mostly anything	socially	NA	graduated from college/university

## OK Cupid data



### Data frames

#### Data frames contain structured data

- > library(rworkshop)
- > download\_data("profiles\_revised.csv") # only needs to be run once
- > profiles <- read.csv("profiles\_revised.csv")
- > View(profiles) # the View() function only works in R Studio!

•	age 🗦	body_type	diet <sup>‡</sup>	drinks <sup>‡</sup>	drugs <sup>‡</sup>	education
1	22	a little extra	strictly anything	socially	never	working on college/university
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## Data Frames

## Variables

	(	1	)
	(	1	)
	ĺ		)
	(	τ	5
(		_	)
1			

•	age 🗦	body_type	diet <sup>‡</sup>	drinks <sup>‡</sup>	drugs <sup>‡</sup>	education
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# An Example Dataset

### **Quantitative Variable**

### **Categorical Variable**

Cases (observational units)

•	age	body_type	diet <sup>‡</sup>	drinks	÷	drugs <sup>‡</sup>	education
1	22	a little extra	strictly anything	socially		never	working on college/university
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## Data frames

We can extract the columns of a data frame as vector objects using the \$ symbol

> the\_ages <- profiles\$age

Can you get the mean() age of users in this data set?

> mean(the\_ages)

# Extracting rows from a data frame

We can extract rows from a data frame in a similar way as extracting values from a vector by using the square brackets

- > profiles[1, ] # returns the first row of the data frame
- > profiles[, 1] # returns the first column of the data

Note, the first column of the profiles data frame is the variable age, so we can also get the first column using:

> profiles\$age # this is the same as profiles[, 1]

# Extracting rows from a data frame

We can also create vectors of numbers or Booleans specifying which rows we want to extract from a data frame

```
# create a vector with the numbers 1, 10, 20 > my vec <- c(1, 10, 20)
```

- # use my\_vec to get the 1<sup>st</sup>, 10<sup>th</sup>, and 20<sup>th</sup> row in profiles
- > small\_profiles <- profiles[my\_vec, ]
- > dim(small\_profiles) # number of rows and columns in the data frame

## Extracting rows from a data frame

Finally, we can also extract rows by creating a Boolean vector that is of the same length as the number of rows in the data frame

TRUE values will be extracted from the data frame while FALSE values will not

```
# create a vector of booleans
> my_bools <- c(TRUE, FALSE, TRUE)

# use the Boolean vector to get the 1<sup>st</sup> and 3<sup>rd</sup> row
> small_profiles[my_bools,]
```

# Questions?



# Break time?

