CSCE 587 Introduction to R / RStudio

Part 2

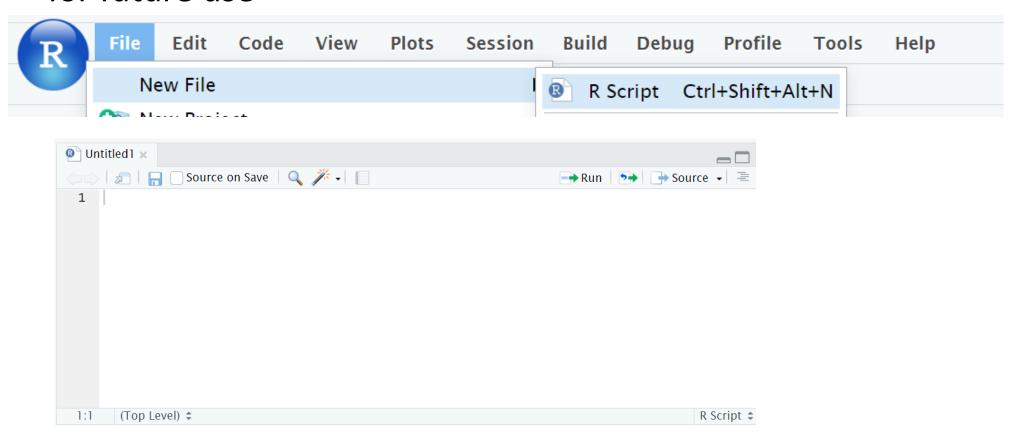
RStudio Environment: Console / History / Script Files

Loading Datasets and Basic Plots

- Console R commands are executed
- Terminal Your Virtual Machine terminal



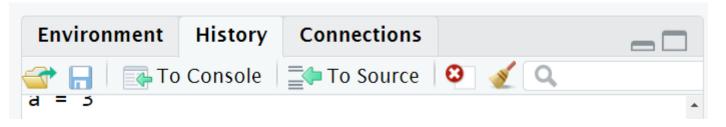
 Editor – We can edit R Script files (.R) to save on our virtual machine for future use



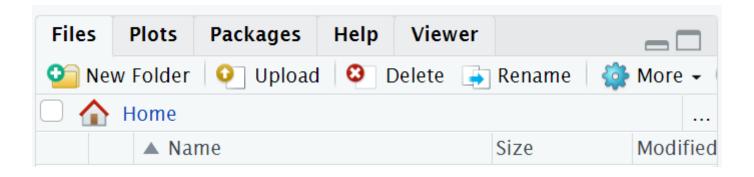
Environment – View objects currently held in the environment



 History – View recently executed commands. Can load, save, execute, save, or remove/clear these commands



Manage Files and Packages, view plots and help

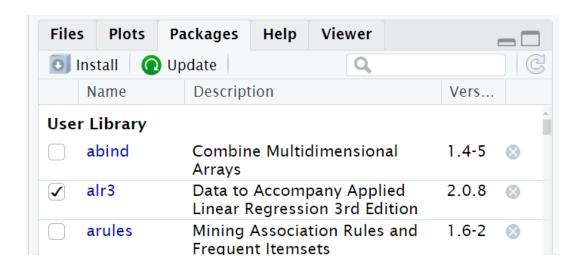


Loading Data Sets

- Several ways to load data sets
 - Data sets in R packages
 - Data sets from the web
 - Data sets on the VM

Loading Data Sets from R Packages

Click on the Packages button
The second package listed is alr3
Click in the box at the left to load the package



Clicking the box in the GUI causes the R command to load the library to be executed at the console.

```
> library("alr3", lib.loc="~/R/x86_64-redhat-linux-gnu-library/3.5")
Loading required package: car
Loading required package: carData
```

Loading Data Sets from R Packages

- We'll use data() to access the UN2 data set in the alr3 package
- > data(UN2)
- > UN2

655347	0.0
00001	22
890338	43
854268	58
479969	35
868811	88
016339	67
655347	91
561438	67
	655347 890338 854268 479969 868811 016339 655347 561438

This data set looks like a matrix with row and column labels Let's examine the first row:

```
> UN2[1,]
```

```
logPPgdp logFertility Purban Afghanistan 6.61471 2.765535 22
```

Note: R shows us the row and column labels in the output

Row and Column labels can be used for subsetting

```
> UN2['Algeria',]
```

```
logPPgdp logFertility Purban Algeria 10.8009 1.485427 58
```

This is better than hard coding an index since the addition of another row to the dataset won't affect which row our command accesses!

We can calculate the mean for each column using the column names

```
> mean(UN2[,'logPPgdp'])
[1] 10.99309
> mean(UN2[,'logFertility'])
[1] 1.468687
> mean(UN2[,'Purban'])
[1] 55.53886
```

Since all the values are numeric, we can compute the variancecovariance:

> var(UN2)

```
logPPgdp logFertility Purban
logPPgdp 5.640839 -1.2475279 44.55587
logFertility -1.247528 0.6009038 -11.00879
Purban 44.555873 -11.0087939 579.19770
```

Example URL for a binary data file:

https://cse.sc.edu/~bhipp/587/WavesBasicR.RData

To download to our virtual machine, select the Terminal window in RStudio and enter:

wget https://cse.sc.edu/~bhipp/587/WavesBasicR.RData --no-check-certificate

Excerpt from the terminal window showing command and result

[student@sandbox-host ~]\$ wget https://cse.sc.edu/~bhipp/587/WavesBasicR.RData --no-check-certificate

--2024-08-25 19:29:08-- https://cse.sc.edu/~bhipp/587/WavesBasicR.RData

Resolving cse.sc.edu (cse.sc.edu)... 129.252.138.13

Connecting to cse.sc.edu (cse.sc.edu) | 129.252.138.13 | :443... connected.

HTTP request sent, awaiting response... 200 OK

Length: 1472151 (1.4M)

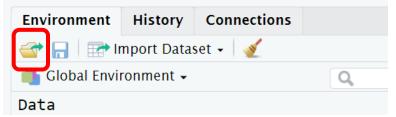
Saving to: 'WavesBasicR.RData'

100%[=======>] 1,472,151 --.-K/s in 0.01s

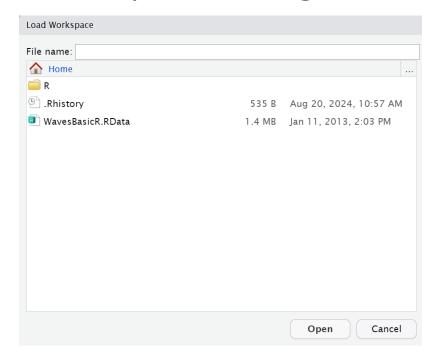
2024-08-25 19:29:08 (102 MB/s) - 'WavesBasicR.RData' saved [1472151/1472151]

Now we can load the data set into RStudio

Click on the load workspace icon



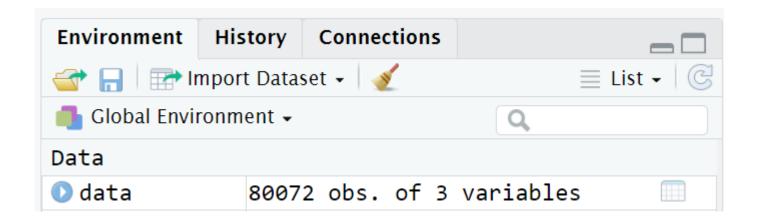
This will open a dialog box for us to select the WavesBasicR.Rdata file



Notice that opening the file through the GUI causes the load command to be execute in the console window:

> load("~/WavesBasicR.RData")

The data set is a data.frame called "data"



A data frame is similar to a matrix, but columns don't have to all be of the same type (though the values within an individual column are all the same type).

The summary function is useful to provide some basic descriptive statistics for numeric fields in a dataframe.

Example:

> summary(data)

```
wave height
                    X
    :-66.15 Min. :-179.998
                                Min. : 3.500
Min.
1st Qu.:-57.96
              1st Qu.:-103.803
                                1st Qu.: 3.917
Median :-50.00
               Median: -2.345
                                Median : 4.462
Mean : -42.71
               Mean : -4.924
                                Mean : 4.841
3rd Qu.:-39.50
               3rd Qu.: 88.488
                                3rd Qu.: 5.327
               Max. : 179.999
                                Max. :19.986
Max. : 66.15
```

Using this data set, how many waves are higher than the mean height? Let's start by looking at the first few rows of data

> data[1:10,]

```
x wave height
  -66.13506 18.82452
                            5.408
30 -66.09778 20.76402
                           5.053
                           4.663
  -66.03464 22.69563
                     4.937
60 -65.94584 24.61539
                           4.653
75 -65.83165 26.51951
                            4.765
   -65.69244 28.40436
105 -65.52863 30.26658
                           4.370
120 -65.34071 32.10304
                            4.299
135 -65.12923 33.91093
                            4.066
150 -64.89478 35.68771
                            3.682
```

```
Note: the third column is the wave height
```

- 1. Start by calculating the mean (use the column label)
- > mean(data[,'wave_height'])

```
[1] 4.841396
```

- 2. Determine "which" observations are larger than the mean?
- Hint: use which() and save the resulting vector
- > larger = which(data[,'wave_height']>mean(data[,'wave_height']))
- 3. Get length of vector. This is our answer!
- > length(larger)

[1] 29662

Plot – From Base R

Let's look at the help first

> ?plot

Plot the waves

> plot(data\$x, data\$y, main="Coordinates of wave heights")

Histograms – From Base R

We can use hist() to plot a histogram

Syntax: hist(data, xlab, main)

> hist(data\$wave_height, xlab="wave heights", main="Our first histogram")

Plotting Histograms

Let's use the ggplot2 package to produce a nicer looking histogram Load the ggplot2 package (using the library function) Use the following command:

> ggplot(data=data, aes(data\$wave_height)) + geom_histogram()

Plotting Histograms

Let's add our own labels and title

```
> ggplot(data=data, aes(data$wave_height)) +
geom_histogram() +
labs(title="Our histogram using ggplot", x="wave height", y="count")
```

Plotting Histograms

Using hist() to plot a histogram can be easier.

> hist(data\$wave_height, xlab="wave heights", main="Our first histogram")

Let's add a yellow dashed vertical line were the mean is in the histogram

> abline(v=mean(data\$wave_height), col="yellow", lwd=3, lty=2)

Plotting Histograms with ggplot2 package

Let's update the color and number of bins

```
> ggplot(data=data, aes(data$wave_height)) +
geom_histogram(bins=30, col="red", fill="blue") +
labs(title="Our ggplot histogram",x="wave height", y="count")
```

Plotting Histograms ggplot2 package

Let's add a yellow dashed vertical line were the mean is in the histogram

```
> ggplot(data=data, aes(data$wave_height)) +
geom_histogram(bins=30, col="red", fill="blue") +
labs(title="Our ggplot histogram",x="wave height", y="count") +
geom_vline(aes(xintercept=mean(data$wave_height)),
color="yellow", linetype="dashed", size=1)
```

Saving Our Plot

Step 1: Open a pdf file using pdf()

Syntax: pdf(filename, width, height)

Step 2: draw the plot

Step 3: close the file using dev.off()

Saving Our Plot

```
Step 1: open the file
> pdf("myHist.pdf", width=6, height=8)
Step 2: Draw the plot
> ggplot(data=data, aes(data$wave height)) +
geom histogram(bins=30, col="red", fill="blue") +
labs(title="Our ggplot histogram",x="wave height", y="count") +
geom vline(aes(xintercept=mean(data$wave height)),
color="yellow", linetype="dashed", size=1)
Step 3: close the file
> dev.off()
```

Boxplots

What is represented in the boxplot?

Q1 = First Quartile / 25th Percentile

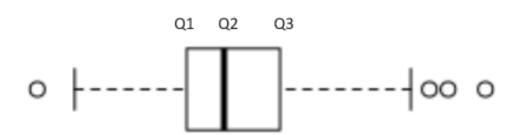
Q2 = Second Quartile / 50th Percentile / Median

Q3 = Third Quartile / 75th Percentile

IQR = Q3 – Q1 = Interquartile Range

Whiskers extend to smallest and largest values in the dataset within 1.5 *IQR of Q1 and Q3

Outliers are values more than 1.5 * IQR below Q1 or above Q3



Boxplots

Use boxplot() to draw a boxplot:

Syntax: boxplot(dataset, x-label, title)

> boxplot(data, xlab= "Variable Names", main= "Boxplot of wave height data")

Try y alone

> boxplot(data\$y, xlab= "Variable Names", main= "Boxplot of wave height data")

Let's plot the wave data again:

> plot(data\$x, data\$y, main="Coordinates of wave heights")

Now, lets emphasis larger waves: use a different color:

Step 1: find waves larger than some threshold

Step 2: keep the indices of those "larger" waves

Step 3: add those larger waves using a different color

Step 1: find waves larger than some threshold. How?

```
> data$wave_height > 7
```

Step 2: keep the indices of those "larger" waves

> ind = which(data\$wave_height > 7)

How many?

> length(ind)

Step 3: Add those larger waves using a different color

Use the points() operation

Syntax: points(x, y, col, symbol)

> points(data\$x[ind], data\$y[ind], col="red", pch=20)

Installing a Package

Install the "maps" package by either selecting "Install Packages..." from the Tools menu, or clicking the Install button on the Packages tab

Note: We only have to install a package once. To use the package, we'll load it with the library function

Add the map to our plot

> map("world", add= TRUE)

Determine if the northern or southern hemisphere has the largest waves.

Step 1: Choose an appropriate threshold

Step 2: Select those waves larger than the threshold

Step 3: Highlight these observations in a plot

Step 4: Determine which hemisphere has the largest waves.

Step 1: Choose an appropriate threshold. How?

- 1) Guess? Could take a while.
- 2) Maybe look at the boxplot?
- > boxplot(data\$wave_height)

Step 2: Select those waves larger than the threshold

> large = which(data\$wave_height > 15)

How many waves are larger than 15?

Step 3: Highlight these observations in a plot

> points(data\$x[large], data\$y[large], col="yellow", pch=20)

Step 4: Determine which hemisphere has the largest waves. Looks like a toss up. Try a larger threshold?