

# LECTURE 5: MEET R

ECON 480 - ECONOMETRICS - FALL 2018

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Ryan Safner

September 12, 2018

Writing and Saving R Code

Objects in R

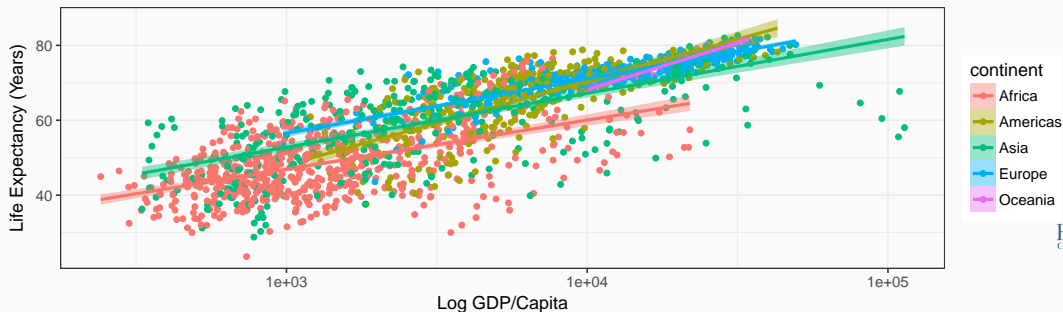
Data Frames

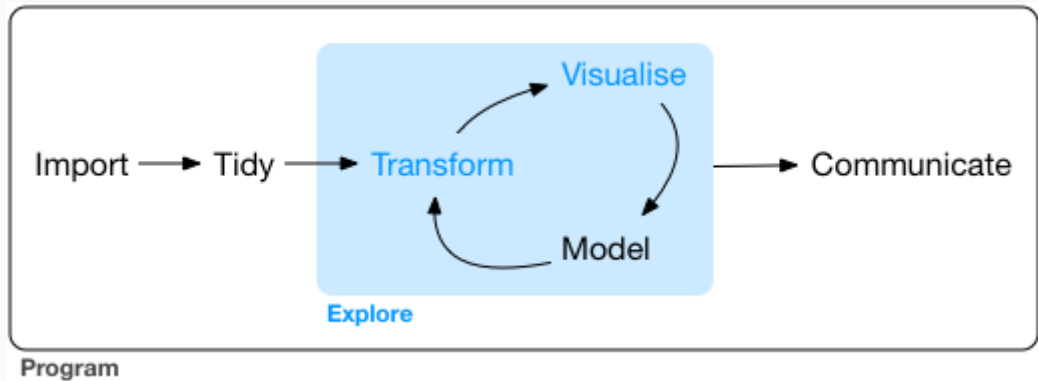
Quick Data Analysis Example

Excel	Stata	R
MS Office License	Expensive License	Free!
Proprietary	Proprietary	Open Source
Default, used in finance	Primary for economists	Largest use by data scientists
Not reproducible	Reproducible (.do files)	Reproducible (.R files)
Can't incorporate into docs	Can't incorporate into docs	Can run in chunks within docs
Very limited extensions	Many packages	Most packages written for R first
Point-and-click	Point-and-click or command line	Almost exclusively command line
Limited formulas	Just one command per task	Many alternative commands

## STATA AND EXCEL CAN'T DO THIS (RIGHT IN THE SLIDES)

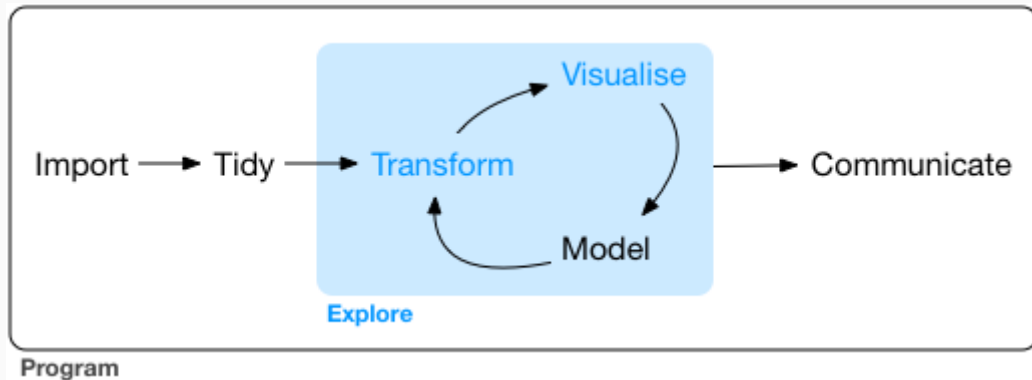
```
ggplot(data = gapminder, aes(x = gdpPercap,  
  y = lifeExp, color = continent, fill= continent))+  
  geom_point()+geom_smooth(method = "lm") +  
  scale_x_log10()+ylab("Life Expectancy (Years)") +  
  xlab("Log GDP/Capita")
```





Workflow

- Need software that can import, tidy, analyze, plot, and present data



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- R can do all of this (with *R Markdown*, all in the same document)

- You are literally learning a new language, complete with grammar and syntax rules, and specific vocabulary

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- R like all command line programming is *very literal*, a single typo or misplaced comma will lead to a different outcome than you intended, or fail completely

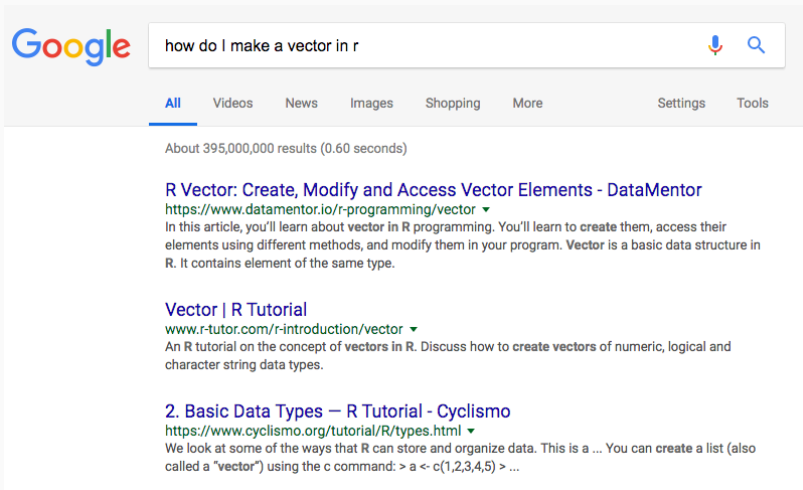


"There's an implied contract between you and R: it will do the tedious computation for you, but in return, you must be completely precise in your instructions. Typos matter. Case matters."

[R for Data Science](#)



Hadley Wickham  
Chief Scientist, R Studio



A screenshot of a Google search interface. The search bar contains the text "how do I make a vector in r". Below the search bar, there are tabs for "All", "Videos", "News", "Images", "Shopping", and "More". The "All" tab is selected. Below the tabs, it says "About 395,000,000 results (0.60 seconds)". There are three search results listed. The first result is titled "R Vector: Create, Modify and Access Vector Elements - DataMentor" with a URL "https://www.datamentor.io/r-programming/vector" and a description: "In this article, you'll learn about vector in R programming. You'll learn to create them, access their elements using different methods, and modify them in your program. Vector is a basic data structure in R. It contains element of the same type." The second result is titled "Vector | R Tutorial" with a URL "www.r-tutor.com/r-introduction/vector" and a description: "An R tutorial on the concept of vectors in R. Discuss how to create vectors of numeric, logical and character string data types." The third result is titled "2. Basic Data Types – R Tutorial - Cyclismo" with a URL "https://www.cyclismo.org/tutorial/R/types.html" and a description: "We look at some of the ways that R can store and organize data. This is a ... You can create a list (also called a 'vector') using the c command: > a <- c(1,2,3,4,5) > ..."

Google

how do I make a vector in r


All Videos News Images Shopping More Settings Tools

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**R Vector: Create, Modify and Access Vector Elements - DataMentor**  
<https://www.datamentor.io/r-programming/vector> ▼  
In this article, you'll learn about vector in R programming. You'll learn to create them, access their elements using different methods, and modify them in your program. Vector is a basic data structure in R. It contains element of the same type.

**Vector | R Tutorial**  
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 **stackoverflow**

QuestionsDeveloper JobsTagsUsers

[r] how do I make a vector

Search

results found containing **how do I make a vector** tagged with **r**

[r] how do I make a vector

search

500 results

relevance

newest

votes

active

**R** is a free, open-source programming language and software environment for statistical computing, bioinformatics, visualization and general computing. Provide minimal, reproducible, representative example(s) with your questions. Use `dput()` for data and specify all non-base packages with `library` ...

[Learn more...](#) [Top users](#) [Synonyms \(2\)](#) [r jobs](#)

2  
votes

2  
answers

**Q: How do I make a specific factor in a vector have a higher level than every other factor?**

Given **a vector** for which "b" will always be an element of, **how do I make** "b" have **a** higher level than all the other factors (without reordering the other factors relative to each other)? For example ... , but **how do I make** it so `levels(df$x) = "c","d","b"` In other words, **I want** "b" to always show up last. ...

**r**

asked Dec 26 '13 by Ben

```
#type help("functionname") to get documentation on the command  
help("lm")
```

lm {stats}

R Documentation

## Fitting Linear Models

### Description

`lm` is used to fit linear models. It can be used to carry out regression, single stratum analysis of variance and analysis of covariance (although [aov](#) may provide a more convenient interface for these).

### Usage

```
lm(formula, data, subset, weights, na.action,  
   method = "qr", model = TRUE, x = FALSE, y = FALSE, qr = TRUE,  
   singular.ok = TRUE, contrasts = NULL, offset, ...)
```

### Arguments

- |                |  |
|----------------|--|
| <b>formula</b> | an object of class " <a href="#">formula</a> " (or one that can be coerced to that class): a symbolic description of the model to be fitted. The details of model specification are given under 'Details'.   |
| <b>data</b>    | an optional data frame, list or environment (or object coercible by <a href="#">as.data.frame</a> to a data frame) containing the variables in the model. If not found in <code>data</code> , the variables are taken from <code>environment(formula)</code> , typically the environment from which <code>lm</code> is called. |
| <b>subset</b>  | an optional vector specifying a subset of observations to be used in the fitting process.  |
| <b>weights</b> | an optional vector of weights to be used in the fitting process. Should be <code>NULL</code> or a numeric vector. If non- <code>NULL</code>  |

## WRITING AND SAVING R CODE

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- It will be wise to adopt some consistent standard for demarcating names:

`i.use.snake.case`

`otherPeopleUseCamelCase`

`some_people_use_underscores`

`And_aFew.People_RENOUNCEconvention`

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  - OS-specific to Windows vs. Mac vs. Linux

- Comment, comment, comment!

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# Run regression of y on x, save as reg1  
reg1<-lm(y~x, data=data) #runs regression  
summary(reg1$coefficients) #prints coefficients
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  - Better yet, ask me about version control and GitHub (later)

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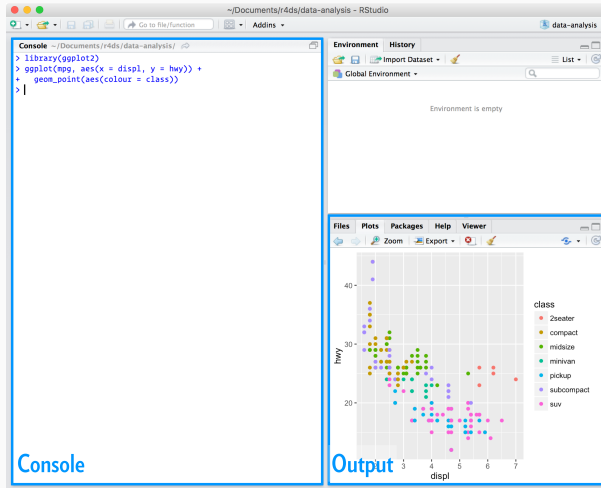
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  - Markdown is a language that is intuitive, simple, human- and machine-readable



Rstudio Windows

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- **Later:** *R Markdown* and the benefits to plain text, literate programming, and workflow management

- First, recognize that R can be used as a simple calculator

```
2+2
```

```
## [1] 4
```

```
sqrt(100/4)
```

```
## [1] 5
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- This will be helpful later when we want to transform variables



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- If you do not have a package, they are easy to install with (yes, note the plural “s”)

```
install.packages("packagename")
```

- Here is a list of the most important packages you will probably use for things relevant to econometrics

Package name	Use
<code>ggplot2</code>	Rendering beautiful graphics (scatterplots, histograms, etc)
<code>stargazer</code>	Rendering professional looking regression output tables
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- We will explore each of these in more detail later



## OBJECTS IN R

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- We can also build vectors via generating series

```
1:5
```

```
## [1] 1 2 3 4 5
```

- We can perform mathematical operations on a vector as a whole:

```
sum(1:5)
```

```
## [1] 15
```

```
mean(1:5)
```

```
## [1] 3
```



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```
list
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```
## [1] "red"    "blue"   "purple"
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```
print(list)
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mixed<-c(5, pi, TRUE, 4.3, "cabbage")  
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- **Logical** is a series of binary elements that can either be TRUE or FALSE

```
logical<-c(TRUE,FALSE,FALSE,TRUE)
```

- **Character** is a string of text: letters, numbers, and symbols, cannot perform mathematical operations

```
character<-c("one","two","7","orange")
```

- **Factor** is a special type of character variable, often used to indicate membership in one of several possible categories, called **levels** (e.g. for plotting, or conditional statistics and data work)

```
students<-factor(c("freshman", "senior", "senior", "junior", "freshman",  
                  "sophomore", "freshman"))
```

```
levels(students) #extract unique levels
```

```
## [1] "freshman" "junior"    "senior"    "sophomore"
```

```
nlevels(students) #count the number of levels
```

```
## [1] 4
```

```
table(students) #tabulate number of values for each level
```

```
## students
##  freshman    junior    senior sophomore
##           3         1         2         1
```

- Again, check the type of data with `class()`

```
class(x)
```

```
## [1] "numeric"
```

```
x<-as.character(x)
```

```
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- Change the type with `as.classname()`

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```
## [1] "character"
```

```
x<-as.numeric(x)
```



- Again, check the type of data with `class()`
- Change the type with `as.classname()`
  - Note you can't turn characters into numeric (if there's no numbers), but you can turn numeric into characters

```
class(x)
```

```
## [1] "numeric"
```

```
x<-as.character(x)
```

```
class(x)
```

```
## [1] "character"
```

```
x<-as.numeric(x)
```

- Use [square brackets] to isolate elements of a vector for commands

```
print(list) #Our original list
```

```
## [1] "red"      "blue"      "purple"
```

```
list[1] #Inspect first element
```

```
## [1] "red"
```

```
list[c(1,3)] #Inspect first and third elements
```

```
## [1] "red"      "purple"
```

- Use [square brackets] to isolate elements of a vector for commands

```
list[2]<-"orange" #Change second element  
print(list) #Our new list
```

```
## [1] "red"      "orange" "purple"
```

- Keeping our first vector  $x$ , let's create another object  $y$

```
y <- 10
```

- Keeping our first vector  $x$ , let's create another object  $y$

```
y<-10
```

- Let's create a third object  $z$ , which is the product of  $x$  and  $y$

```
z<-x*y
```

```
z
```

```
## [1] 10 20 30 40 50
```

- Objects and variables maintain their value until they are changed. We can redefine `x` as 6 simply with another command to define `x`.

```
x
```

```
## [1] 1 2 3 4 5
```

```
x<-6
```

```
x
```

```
## [1] 6
```

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- A **function** produces a (hopefully useful) output based on the input that it receives.



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- Most of R: `functionname(objectname)`
- A **function** produces a (hopefully useful) output based on the input that it receives.
- Functions can be recognized by the parentheses () at the end of their names.
- The `c()` command that produces a vector, was an example of a function.

- Mathematical/Statistical Functions

Function	Use	Example
<code>sum()</code>	Takes the sum of an object	<code>sum(1:10)</code>
<code>mean()</code>	Takes the arithmetic mean of an object	<code>mean(1:10)</code>
<code>rnorm()</code>	Takes a number of draws (e.g. 5) from a normal distribution	<code>rnorm(5)</code>
<code>round()</code>	Rounds an object (e.g. x) to (e.g. 2) decimal places	<code>round(x,2)</code>

## SOME USEFUL FUNCTIONS

- Mathematical/Statistical Functions

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<code>round()</code>	Rounds an object (e.g. x) to (e.g. 2) decimal places	<code>round(x,2)</code>

- Note functions can have functions in their arguments, e.g. `round(rnorm(5),2)`

## DATA FRAMES

---

- The most important object in R is a **data frame**, used for statistics, plots, regressions, etc

```
gapminder
```

```
## # A tibble: 1,704 x 6
```

```
##   country      continent  year lifeExp      pop gdpPercap
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl>
## 1 Afghanistan Asia      1952   28.8  8425333    779.
## 2 Afghanistan Asia      1957   30.3  9240934    821.
## 3 Afghanistan Asia      1962   32.0 10267083    853.
## 4 Afghanistan Asia      1967   34.0 11537966    836.
## 5 Afghanistan Asia      1972   36.1 13079460    740.
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- The most important object in R is a **data frame**, used for statistics, plots, regressions, etc
  - “Rectangular” data (i.e. “spreadsheet”): rows are observations, columns are variables
  - Can hold variables of different classes
  - All vectors (columns) must have the same length!

gapminder

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```

```
##   country      continent  year lifeExp      pop gdpPercap
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```

- We often import existing data into a dataframe (see importing data below)

```
v1<-c(10,20,30,45,60) # A list of integers
v2<-LETTERS[1:5] # The first 5 letters
v3<-round(rnorm(5),2) #5 random draws from normal distr, rounded to 2 decimal pl
v4<-c("Apples","Oranges","Bananas","Kiwi","Watermelon") #Fruits
mydf<-data.frame(v1,v2,v3,v4) #make dataframe called mydf
```

- We often import existing data into a dataframe (see importing data below)
- We could construct a data frame from scratch using `data.frame()`

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v1<-c(10,20,30,45,60) # A list of integers
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```

- We often import existing data into a dataframe (see importing data below)
- We could construct a data frame from scratch using `data.frame()`
- Suppose we have 4 different vectors, `v1`, `v2`, `v3`, and `v4`

```
v1<-c(10,20,30,45,60) # A list of integers
v2<-LETTERS[1:5] # The first 5 letters
v3<-round(rnorm(5),2) #5 random draws from normal distr, rounded to 2 decimal places
v4<-c("Apples","Oranges","Bananas","Kiwi","Watermelon") #Fruits
mydf<-data.frame(v1,v2,v3,v4) #make dataframe called mydf
```

- Check the structure of a data frame with `str()`

```
str(mydf) #examine structure
```

```
## 'data.frame':    5 obs. of  4 variables:  
## $ v1: num  10 20 30 45 60  
## $ v2: Factor w/ 5 levels "A","B","C","D",...: 1 2 3 4 5  
## $ v3: num  0.58 0.19 0.1 -0.14 0.81  
## $ v4: Factor w/ 5 levels "Apples","Bananas",...: 1 4 2 3 5
```

```
class(mydf) #check it's a dataframe
```

```
## [1] "data.frame"
```

- Note instead of making the vectors first and then combining them, we could have done it all at once:

```
mydf<-data.frame(v1=c(10,20,30,45,60),  
                 v2=LETTERS[1:5],  
                 v3=round(rnorm(5),2),  
                 v4=c("Apples","Oranges","Bananas","Kiwi","Watermelon"))
```

- We will use the `gapminder` dataset as a quick example

```
library("gapminder")  
str(gapminder)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':    1704 obs. of  6 variables:  
## $ country   : Factor w/ 142 levels "Afghanistan",...: 1 1 1 1 1 1 1 1 1 1 ...  
## $ continent: Factor w/ 5 levels "Africa","Americas",...: 3 3 3 3 3 3 3 3 3 3 ...  
## $ year      : int   1952 1957 1962 1967 1972 1977 1982 1987 1992 1997 ...  
## $ lifeExp   : num   28.8 30.3 32 34 36.1 ...  
## $ pop       : int   8425333 9240934 10267083 11537966 13079460 14880372 16811859 19242662 21324954 23214761 ...  
## $ gdpPercap: num    779 821 853 836 740 ...
```

## DATA FRAMES: QUICK LOOK

- We will use the `gapminder` dataset as a quick example
- `str()` will give us a sense of the structure

```
library("gapminder")
```

```
str(gapminder)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':    1704 obs. of  6 variables:
##  $ country   : Factor w/ 142 levels "Afghanistan",...: 1 1 1 1 1 1 1 1 1 1 ...
##  $ continent: Factor w/ 5 levels "Africa","Americas",...: 3 3 3 3 3 3 3 3 3 3 ...
##  $ year      : int   1952 1957 1962 1967 1972 1977 1982 1987 1992 1997 ...
##  $ lifeExp   : num   28.8 30.3 32 34 36.1 ...
##  $ pop       : int   8425333 9240934 10267083 11537966 13079460 14880372 16811859 19222675 21324554 23214761 ...
##  $ gdpPercap: num    779 821 853 836 740 ...
```



- `head()` will show us the top 6 rows (observations)

```
head(gapminder)
```

```
## # A tibble: 6 x 6
##   country      continent  year lifeExp      pop gdpPercap
##   <fct>        <fct>    <int>  <dbl>    <int>    <dbl>
## 1 Afghanistan Asia      1952   28.8  8425333    779.
## 2 Afghanistan Asia      1957   30.3  9240934    821.
## 3 Afghanistan Asia      1962   32.0 10267083    853.
## 4 Afghanistan Asia      1967   34.0 11537966    836.
## 5 Afghanistan Asia      1972   36.1 13079460    740.
## 6 Afghanistan Asia      1977   38.4 14880372    786.
```

- `summary()` will give us a summary statistics of each variable (columns)

```
summary(gapminder)
```

```
##           country      continent      year      lifeExp
## Afghanistan: 12 Africa :624 Min.      :1952 Min.      :23.60
## Albania      : 12 Americas:300 1st Qu.:1966 1st Qu.:48.20
## Algeria       : 12 Asia     :396 Median  :1980 Median  :60.71
## Angola        : 12 Europe  :360 Mean     :1980 Mean     :59.47
## Argentina     : 12 Oceania : 24 3rd Qu.:1993 3rd Qu.:70.85
## Australia     : 12              Max.     :2007 Max.     :82.60
## (Other)       :1632
##           pop      gdpPercap
## Min.      :6.001e+04 Min.      : 241.2
## 1st Qu.: 2.531e+05 1st Qu.: 1023.1
```

- Each variable is stored as a part of a data frame that can be called with the \$ sign

## DATA FRAMES: SELECTING COLUMNS (VARIABLES)

- Each variable is stored as a part of a data frame that can be called with the \$ sign
  - e.g. with the Diamonds data, price can be called with `Diamonds$price`:

```
summary(gapminder$gdpPercap)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
##  241.2    1202.1    3531.8    7215.3    9325.5   113523.1
```

```
mean(gapminder$gdpPercap)
```

```
## [1] 7215.327
```

## QUICK DATA ANALYSIS EXAMPLE

---

- Simple commands can help us learn about a dataset quickly

Function	Result
<code>min(distr)</code>	Find minimum value
<code>max(distr)</code>	Find maximum value
<code>range(distr)</code>	Find the range
<code>sort(distr)</code>	Sort values of distribution from smallest to largest
<code>sort(distr)[1]</code>	Find first value when sorted (equivalent to finding min)
<code>sort(distr, decreasing=TRUE)</code>	Sort from largest to smallest
<code>median(distr)</code>	Find the median
<code>mean(distr)</code>	Find the mean
<code>var(distr)</code>	Find the variance
<code>sd(distr)</code>	Find the standard deviation

Function	Result
<code>table(distr)</code>	Gives frequency table of categorical variable values
<code>fivenum(distr)</code>	Five number summary (min, q1, median, q3, max)
<code>summary(distr)</code>	Gives min, q1, median, mean, q3, max
<code>quantile(distr, 0.32)</code>	Find specific (e.g. 32nd) percentile
<code>summary(factor(distr))</code>	Lists all unique values in distr
<code>sum(distr)</code>	Takes the sum of all values in distr

```
summary(gapminder$gdpPercap)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
##    241.2   1202.1   3531.8   7215.3   9325.5  113523.1
```

```
mean(gapminder$pop)
```

```
## [1] 29601212
```

```
table(gapminder$continent)
```

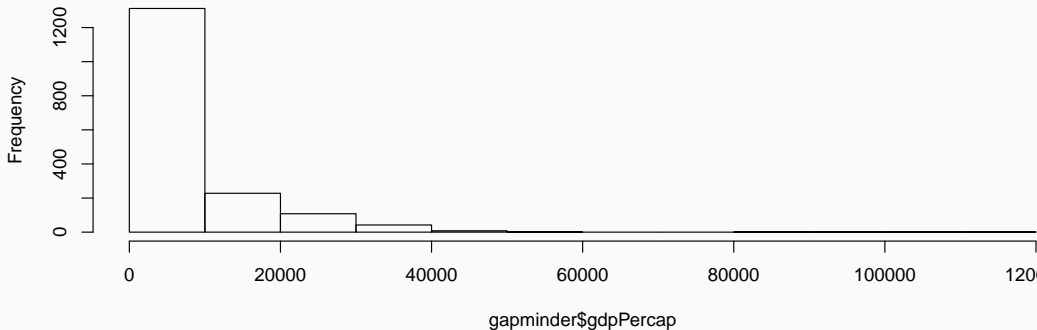
```
##
## Africa Americas      Asia  Europe  Oceania
##    624     300     396     360     24
```



- Base R is very powerful and intuitive to plot, but not very sexy

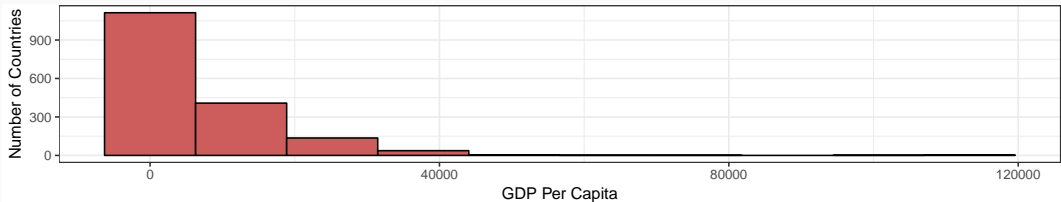
```
hist(gapminder$gdpPercap)
```

**Histogram of gapminder\$gdpPercap**



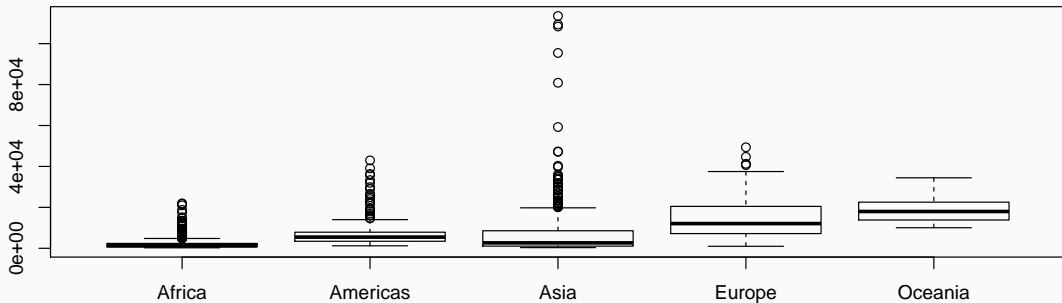
- This is where packages (like `ggplot2`) come in, but we'll have to learn later

```
library("ggplot2")  
ggplot(gapminder, aes(x=gdpPercap))+  
  geom_histogram(bins=10, color="black", fill="indianred")+  
  xlab("GDP Per Capita")+ylab("Number of Countries")+  
  theme_bw()
```



- Boxplots

```
boxplot(gdpPercap~continent,data=gapminder)
```



## PLOTTING IV

```
library("ggplot2")  
library("gapminder")  
ggplot(gapminder, aes(x=continent,y=gdpPercap ,fill=continent))+  
  geom_boxplot()+ theme_bw()
```

