

Data Analytics and Visualizations in R - Exercises

Contents

1	Basic R Data Structures	2
1.1	Types	2
1.2	Weirdness of R	2
1.3	Atomic Vector Concatenation	3
1.4	Vector Concatenation	3
1.5	From Vectors to <code>data.frames</code>	4
1.6	Attributes	6
1.7	Factors	8
1.8	More fun with factors	8
1.9	Creating <code>data.frames</code>	9
1.10	Combining <code>data.frames</code>	9
1.11	Computation on <code>data.frames</code>	10
1.12	Missing Values	13
2	Advanced R Data Structures and Mathematical Operations	13
2.1	Basic data structures	13
2.2	14
2.3	14
2.4	14
2.5	14
2.6	<code>lapply()</code>	15
2.7	15
2.8	16
2.9	16
2.10	16
2.11	16
2.12	16
2.13	16
2.14	17
2.15	17
3	Data Import	17
3.1	Flat File - Q1	17
3.2	Flat File - Q2	17
3.3	Flat File - Q3	18
3.4	Flat File - Q4	19
3.5	Excel Questions - Q1	20
3.6	Excel Question - Q2	20
3.7	Excel Question - Q3	20
3.8	Excel Question - Q4	20
3.9	Excel Questions	20
3.10	XML - Q1	20
3.11	XML - Q2	20
3.12	JSON - Q1	20
3.13	JSON - Q2	20
3.14	SQL - Q1	21
4	Grammar of graphics and plotting I	21

4.1	Setup	21
4.2	Q1	21
4.3	Q2	21
4.4	Q3	22
4.5	Q4	34
5	Datacamp	35
5.1	ggplot2 on Datacamp	35
5.2	Exploring ggplot2, part 3	37
5.2.1	Instructions	37
5.3	Understanding Variables	40
5.4	Exploring ggplot2, part 4	40
5.4.1	Instructions	41
5.5	Exploring ggplot2, part 5	42
5.5.1	Instructions	42

1 Basic R Data Structures

1.1 Types

What are the scalar types in R?

```
# Scalars are just vectors of length one
```

1.2 Weirdness of R

What is the major difference between atomic vectors and lists? How can you turn a list into an atomic vector?

In order to check if an object is of a certain type you can use `is.[type](object)` ,e.g. `is.integer(object)`

Can you use the `is.vector()` function to understand whether a data structure is a vector? If not, what are the functions that you can use for this purpose?

```
# Atomic vectors can contain elements of the same type. The elements of a list  
# can have different types.
```

```
# Yes, we can use the `is.vector()` function to understand is the structure is a  
# vector. It will return TRUE or FALSE.
```

```
a <- list(c(1,2,3), "bla", TRUE)  
print(a)
```

```
## [[1]]  
## [1] 1 2 3  
##  
## [[2]]  
## [1] "bla"  
##  
## [[3]]  
## [1] TRUE
```

```
is.vector(a)
```

```
## [1] TRUE
```

```
# Best response: If possible you should use type specific coercions like  
# `as.numeric()` or `as.character()`. But since lists are heterogenous, this  
# might not work. A more general function is `unlist()`, which returns the list  
# into a vector of the most general type. Notice this difference:
```

```
a <- list(c(1,2,3), "bla", TRUE)  
unlist(a)
```

```
## [1] "1"      "2"      "3"      "bla"    "TRUE"
```

```
as.character(a)
```

```
## [1] "c(1, 2, 3)" "bla"      "TRUE"
```

```
# Generally you can, but here comes the weird part: `is.vector()` will only return `TRUE` if the vector  
# has no attributes `names`. Therefore more specific functions like `is.atomic()` or `is.list()` functions  
# can be used to test if an object is actually atomic vector or a list
```

1.3 Atomic Vector Concatenation

What happens when you try to generate an atomic vector with `c()` which is composed of different types of elements? What is the `mean()` of a logical vector?

```
# When we attempt to combine different types they will be coerced to the most  
# flexible type. Types from least to most flexible are: logical, integer, double  
# and character.
```

```
str(c("a", 1)) # 1 coerced to char
```

```
## chr [1:2] "a" "1"
```

```
# As TRUE is encoded as 1 and FALSE as 0, the mean is the number of TRUEs  
# divided by the vector length.
```

```
mean(c(TRUE, FALSE, FALSE))
```

```
## [1] 0.3333333
```

1.4 Vector Concatenation

Compare X and Y where X and Y are defined as follows. What is the difference?

```
x <- list(list(1,2), c(3,4))  
y <- c(list(1,2), c(3,4))
```

```
# Answer  
# X will combine several lists into one. Given a combination of atomic vectors  
# and lists, y will coerce the vectors to lists before combining them.  
str(x)
```

```
## List of 2  
## $ :List of 2  
## ..$ : num 1  
## ..$ : num 2
```

```
## $ : num [1:2] 3 4
```

```
str(y)
```

```
## List of 4
## $ : num 1
## $ : num 2
## $ : num 3
## $ : num 4
```

1.5 From Vectors to data.frames

First, create three named numeric vectors of size 10, 11 and 12 respectively in the following manner:

- One vector with the “colon” approach: *from:to*
- One vector with the `seq()` function: *seq(from, to)*
- And one vector with the `seq()` function and the *by* argument: *seq(from, to, by)*

For easier naming you can use the vector `letters` or `LETTERS` which contain the latin alphabet in small and capital, respectively. In order to select specific letters just use e.g. `letters[1:4]` to get the first four letters. Check their types. What is the outcome? Where do you think does the difference come from?

Then combine all three vectors in a list. Check the attributes of the vectors and the list. What is the difference and why?

Finally coerce the list to a `data.frame` with `as.data.frame()`. Why does it fail and how can we fix it? What happened to the names?

Hint: If list elements have no names, we can access them with the double brackets and an index, e.g. `my_list[[1]]`

```
# Answer
```

```
# A. create vectors
```

```
aa <- 1:10
names(aa) <- letters[aa]
aa
```

```
## a b c d e f g h i j
## 1 2 3 4 5 6 7 8 9 10
```

```
bb <- seq(1, 11)
names(bb) <- letters[bb]
bb
```

```
## a b c d e f g h i j k
## 1 2 3 4 5 6 7 8 9 10 11
```

```
cc <- seq(1, 12, by=1)
names(cc) <- letters[cc]
```

```
typeof(aa)
```

```
## [1] "integer"
```

```
typeof(bb)
```

```
## [1] "integer"
```

```
typeof(cc)
```

```
## [1] "double"
```

```
# B. Combine all three vectors in a list
```

```
my_list <- list(aa, bb, cc)
```

```
my_list
```

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

```
##  a b c d e f g h i j
```

```
##  1 2 3 4 5 6 7 8 9 10
```

```
##
```

```
## [[2]]
```

```
##  a b c d e f g h i j k
```

```
##  1 2 3 4 5 6 7 8 9 10 11
```

```
##
```

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

```
##  a b c d e f g h i j k l
```

```
##  1 2 3 4 5 6 7 8 9 10 11 12
```

```
attributes(aa)
```

```
## $names
```

```
##  [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j"
```

```
attributes(bb)
```

```
## $names
```

```
##  [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k"
```

```
attributes(cc)
```

```
## $names
```

```
##  [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l"
```

```
attributes(my_list)
```

```
## NULL
```

```
# C. Coerce to data.frames
```

```
# my_df <- as.data.frame(my_list)# fails
```

```
# Fixing the length
```

```
my_list[[1]] <- c(my_list[[1]], NA, NA)
```

```
my_list[[2]] <- c(my_list[[2]], NA)
```

```
my_df <- as.data.frame(my_list)
```

```
names(my_df) <- LETTERS[1:3]
```

```
my_df
```

```
##    A B C
```

```
## a  1 1 1
```

```
## b  2 2 2
```

```
## c  3 3 3
```

```
## d  4 4 4
```

```
## e  5 5 5
```

```
## f  6 6 6
```

```
## g  7  7  7
## h  8  8  8
## i  9  9  9
## j 10 10 10
## k NA 11 11
##   NA NA 12
```

1.6 Attributes

Take again our `data.frame` from Question 5.

- Change the row names and the column names of the `data.frame` to capital letters (or small letters, if they are already capital).
- Change the `class` attribute to `list`. What happens?
- Change it now to any name you like. What happens now? What happens if you remove the class attribute

```
# Answer
# A. One possible way through attributes
```

```
attributes(my_df)
```

```
## $names
## [1] "A" "B" "C"
##
## $row.names
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" ""
##
## $class
## [1] "data.frame"
```

```
attr(my_df, "names") <- letters[1:3]
attr(my_df, "row.names") <- LETTERS[1:12]
my_df
```

```
##   a  b  c
## A  1  1  1
## B  2  2  2
## C  3  3  3
## D  4  4  4
## E  5  5  5
## F  6  6  6
## G  7  7  7
## H  8  8  8
## I  9  9  9
## J 10 10 10
## K NA 11 11
## L NA NA 12
```

```
# Or through accessor functions
```

```
names(my_df) <- LETTERS[1:3]
row.names(my_df) <- letters[1:12]
my_df
```

```
##   A  B  C
```

```
## a 1 1 1
## b 2 2 2
## c 3 3 3
## d 4 4 4
## e 5 5 5
## f 6 6 6
## g 7 7 7
## h 8 8 8
## i 9 9 9
## j 10 10 10
## k NA 11 11
## l NA NA 12
```

```
# B.
```

```
attr(my_df, "class") <- "list"
my_df
```

```
## $A
## [1] 1 2 3 4 5 6 7 8 9 10 NA NA
##
## $B
## [1] 1 2 3 4 5 6 7 8 9 10 11 NA
##
## $C
## [1] 1 2 3 4 5 6 7 8 9 10 11 12
##
## attr("row.names")
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l"
## attr("class")
## [1] "list"
```

```
# Answer - the data.frame coerced to a list
```

```
# C
```

```
attr(my_df, "class") <- "Batman"
my_df
```

```
## $A
## [1] 1 2 3 4 5 6 7 8 9 10 NA NA
##
## $B
## [1] 1 2 3 4 5 6 7 8 9 10 11 NA
##
## $C
## [1] 1 2 3 4 5 6 7 8 9 10 11 12
##
## attr("row.names")
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l"
## attr("class")
## [1] "Batman"
```

```
# Answer - Nothing changes
```

1.7 Factors

- What is the difference between a Factor and a Vector?
- Create a vector of length 30 with three levels *Rita Repulsa*, *Lord Zedd* and *Rito Revolto* and equal length for each level
- What happens if you replace the second element of the vector with *Shredder*

```
# Answer
# A. A factor is a vector that can contain only predefined values, and is used
# to store categorical data. It is stored as an integer with a character string
# associated with each integer value

# B.

x <- gl(n=3, k=10, length=30, labels=c("Rita Repulsa", "Lord Zedd", "Rito Revolto"))
str(x)
```

```
## Factor w/ 3 levels "Rita Repulsa",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
levels(x)
```

```
## [1] "Rita Repulsa" "Lord Zedd" "Rito Revolto"
```

```
attributes(x)
```

```
## $levels
```

```
## [1] "Rita Repulsa" "Lord Zedd" "Rito Revolto"
```

```
##
```

```
## $class
```

```
## [1] "factor"
```

```
# C
```

```
x[2] <- "Shredder"
```

```
## Warning in `[<-factor`(`*tmp*`, 2, value = "Shredder"): invalid factor
```

```
## level, NA generated
```

```
# It doesn't work. We get the error 'NA generated'
```

1.8 More fun with factors

```
f1 <- factor(letters)
levels(f1) <- rev(levels(f1))
f2 <- rev(factor(letters))
f3 <- factor(letters, levels = rev(letters))
```

The function `rev` reverses the order of an orderable object. What is the difference between `f1`, `f2` and `f3`? Why?

```
# Answer
```

```
f1 <- factor(letters)
```

```
levels(f1) <- rev(levels(f1))
```

```
# f1 goes from a to z and when we apply the levels(f1), z will become 1 and a=26
```

```
f2 <- rev(factor(letters))
```



```

# f2 goes from z to a. but the levels are not changed.

f3 <- factor(letters, levels = rev(letters))

# f3 goes from a - z, but the underlying encoding goes from z = 1 to a = 26.
# We create the vector with the letters a to z BUT the mapped integer structure
# 26 to 1. Hence the levels but not the vector are reversed

f3

## [1] a b c d e f g h i j k l m n o p q r s t u v w x y z
## Levels: z y x w v u t s r q p o n m l k j i h g f e d c b a

# Reversing f3 will give f1

rev(f3)

## [1] z y x w v u t s r q p o n m l k j i h g f e d c b a
## Levels: z y x w v u t s r q p o n m l k j i h g f e d c b a

```

1.9 Creating data.frames

Create a data.frame with 26 rows like this: Only the first and the last six rows are shown. Hint: Instead of the workaround with list you can also use simply `data.frame(column_name = column_vector, ...)`

```

aa <- seq(1:26)
bb <- seq(from=4, to=4*26, by=4)
cc <- rep(seq(1, 26, 2), each=2)
df <- data.frame(V1 = aa, V2 = bb, V3 = letters[cc])
head(df)

##   V1 V2 V3
## 1  1  4  a
## 2  2  8  a
## 3  3 12  c
## 4  4 16  c
## 5  5 20  e
## 6  6 24  e

tail(df)

##   V1 V2 V3
## 21 21 84  u
## 22 22 88  u
## 23 23 92  w
## 24 24 96  w
## 25 25 100 y
## 26 26 104 y

```

1.10 Combining data.frames

Now take the previous data.frame from Question 10 and reproduce the following `data.frame`. Only the first and the last six rows are shown **Hint:** In order to combine to data.frames by column you can use `cbind(df1, df2, ...)`

```

help(cbind)

df[,1] <- NULL
dd <- rev(rep(seq(1, 26, 2), each = 2))
ee <- seq(0, 1.6, length.out = 26)
df2 <- data.frame(V4 = dd, V5 = ee)
binded_df <- cbind(df2, df)
head(binded_df)

```

```

##   V4    V5 V2 V3
## 1 25 0.000  4  a
## 2 25 0.064  8  a
## 3 23 0.128 12  c
## 4 23 0.192 16  c
## 5 21 0.256 20  e
## 6 21 0.320 24  e

```

```
tail(binded_df)
```

```

##   V4    V5 V2 V3
## 21  5 1.280 84  u
## 22  5 1.344 88  u
## 23  3 1.408 92  w
## 24  3 1.472 96  w
## 25  1 1.536 100 y
## 26  1 1.600 104 y

```

1.11 Computation on data.frames

Create the data.frame *df* with `df <- as.data.frame(matrix(runif(9e6), 3e3, 3e3))` This will create a data.frame with 3000 columns and rows and a total of 9mil values.

Now compute the sum of any row, then compute the sum of any column. Measure the time for both operations. Why are the times different, although the size is the same?

- **Hint1:** The time is measured with the function `system.time(my_function_call())`, e.g: `system.time(mean(my_vector))`
- **Hint2:** The sum can be computed with the sum function `sum(my_vector)`
- **Hint2:** Columns and rows are selected by single brackets. Rows: `df[row_number,]`, Columns: `df[,col_number]`

```

# Answer

df <- as.data.frame(matrix(runif(9e6), 3e3, 3e3))

```

```

# rows
system.time(res <- sum(df[1,]))

```

```

##   user  system elapsed
## 0.023   0.001   0.024
res

```

```
## [1] 1492.189
```

```

# columns
system.time(res2 <- sum(df[,1]))

```

```
##      user  system elapsed
##         0         0         0
```

```
res2
```

```
## [1] 1503.822
```

```
# Look at the structure of the objects over which we are computing the sum
```

```
# Column
```

```
str(df[1,])
```

```
## 'data.frame':    1 obs. of  3000 variables:
```

```
## $ V1 : num 0.514
## $ V2 : num 0.756
## $ V3 : num 0.348
## $ V4 : num 0.881
## $ V5 : num 0.82
## $ V6 : num 0.701
## $ V7 : num 0.261
## $ V8 : num 0.658
## $ V9 : num 0.757
## $ V10 : num 0.465
## $ V11 : num 0.777
## $ V12 : num 0.903
## $ V13 : num 0.681
## $ V14 : num 0.499
## $ V15 : num 0.997
## $ V16 : num 0.672
## $ V17 : num 0.21
## $ V18 : num 0.597
## $ V19 : num 0.347
## $ V20 : num 0.595
## $ V21 : num 0.116
## $ V22 : num 0.194
## $ V23 : num 0.23
## $ V24 : num 0.677
## $ V25 : num 0.443
## $ V26 : num 0.303
## $ V27 : num 0.183
## $ V28 : num 0.869
## $ V29 : num 0.188
## $ V30 : num 0.158
## $ V31 : num 0.401
## $ V32 : num 0.634
## $ V33 : num 0.461
## $ V34 : num 0.943
## $ V35 : num 0.395
## $ V36 : num 0.1
## $ V37 : num 0.679
## $ V38 : num 0.0336
## $ V39 : num 0.996
## $ V40 : num 0.771
## $ V41 : num 0.235
## $ V42 : num 0.318
## $ V43 : num 0.945
## $ V44 : num 0.231
```

```
## $ V45 : num 0.905
## $ V46 : num 0.763
## $ V47 : num 0.187
## $ V48 : num 0.409
## $ V49 : num 0.502
## $ V50 : num 0.553
## $ V51 : num 0.921
## $ V52 : num 0.458
## $ V53 : num 0.467
## $ V54 : num 0.803
## $ V55 : num 0.293
## $ V56 : num 0.259
## $ V57 : num 0.21
## $ V58 : num 0.973
## $ V59 : num 0.081
## $ V60 : num 0.842
## $ V61 : num 0.561
## $ V62 : num 0.0436
## $ V63 : num 0.166
## $ V64 : num 0.626
## $ V65 : num 0.733
## $ V66 : num 0.0732
## $ V67 : num 0.307
## $ V68 : num 0.779
## $ V69 : num 0.0876
## $ V70 : num 0.199
## $ V71 : num 0.435
## $ V72 : num 0.316
## $ V73 : num 0.427
## $ V74 : num 0.898
## $ V75 : num 0.523
## $ V76 : num 0.192
## $ V77 : num 0.0784
## $ V78 : num 0.292
## $ V79 : num 0.0597
## $ V80 : num 0.927
## $ V81 : num 0.118
## $ V82 : num 0.65
## $ V83 : num 0.698
## $ V84 : num 0.879
## $ V85 : num 0.348
## $ V86 : num 0.545
## $ V87 : num 0.137
## $ V88 : num 0.0729
## $ V89 : num 0.368
## $ V90 : num 0.902
## $ V91 : num 0.0657
## $ V92 : num 0.568
## $ V93 : num 0.29
## $ V94 : num 0.68
## $ V95 : num 0.919
## $ V96 : num 0.447
## $ V97 : num 0.524
## $ V98 : num 0.147
```

```
## $ V99 : num 0.525
## [list output truncated]

# Row
str(df[,1])

## num [1:3000] 0.514 0.314 0.903 0.262 0.131 ...

# As we can see the extracted column is a numeric vector. But the extracted
# row is a list. Under the hood the sum function is iterating in C/Fortran
# over the specific structure. Iterating over a native array of doubles is
# faster, than iterating over a structure, where at each position, the value
# has to be retrieved from an object possibly stored somewhere further away
# in memory.
```

1.12 Missing Values

- If NA is just a placeholder for a missing value of the same type and Infinity is of type double, why is Infinity plus NA not Infinity?

Hint:

```
paste(paste(rep((Inf - Inf), 20), collapse = ""), "Batman!")

## [1] "NaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaNNaN Batman!"

# Answer:
# Because infinity is not a well defined number, but a concept with
# the type 'double' in R. Adding or subtracting any number from infinity
# will give infinity. However NA is placeholder for any value of the same
# type and therefore also for infinity. As infinity plus/minus infinity
# is not defined, adding NA to infinity can theoretically lead to
# Nan (not a number). Therefore Inf + NA will produce NA.

Inf + NA

## [1] NA

Inf + 1

## [1] Inf

Inf - Inf

## [1] NaN
```

2 Advanced R Data Structures and Mathematical Operations

2.1 Basic data structures

How would you create a 3 by 4 matrix that contains the numbers 1 to 12 and then convert it into a data frame?

```
# Answer

x <- matrix(1:12, 3,4)
```

```
x <- as.data.frame(x)
x
```

```
##   V1 V2 V3 V4
## 1  1  4  7 10
## 2  2  5  8 11
## 3  3  6  9 12
```

2.2

Please use the data frame you created in the first question for the next 5 questions. How would you select the second row elements at second and fourth column ?

```
x <- data.frame(matrix(1:12, 3, 4))
x[2, c(2,4)]
```

```
##   X2 X4
## 2  5 11
```

2.3

How would you assign zero to the elements at row 2 which are greater than 4?

```
x <- data.frame(matrix(1:12, 3, 4))
x[2, x[2,]>4] <- 0
x
```

```
##   X1 X2 X3 X4
## 1  1  4  7 10
## 2  2  0  0  0
## 3  3  6  9 12
```

2.4

How do you set the rownames to “row1”, “row2”, “row3” and column names to “col1”, “col2”, “col3” and “col4” ? (hint:use function “paste0”)

```
x <- data.frame(matrix(1:12, 3, 4))
rownames(x) <- paste0("row", 1:3)
colnames(x) <- paste0("col", 1:4)
x
```

```
##      col1 col2 col3 col4
## row1    1    4    7   10
## row2    2    5    8   11
## row3    3    6    9   12
```

2.5

How do you assign 0 to all elements in columns “col3” and “col4” by using paste0 function ?

```
x <- data.frame(matrix(1:12, 3, 4))
colnames(x) <- paste0("col", 1:4)
x[, paste0(0, 3:4)] <- 0
x
```

```
##   col1 col2 col3 col4 03 04
## 1    1    4    7   10  0  0
## 2    2    5    8   11  0  0
## 3    3    6    9   12  0  0
```

2.6 lapply()

How do you get the numbers whose mod 2 is 0

- by using lapply() function
- by subsetting the data frame directly?

```
x <- data.frame(matrix(1:12), 3, 4)

lapply(x, function(a) a[(a %% 2) == 0])
```

```
## $matrix.1.12.
## [1] 2 4 6 8 10 12
##
## $X3
## numeric(0)
##
## $X4
## [1] 4 4 4 4 4 4 4 4 4 4 4 4
```

```
x[x %% 2 == 0]
```

```
## [1] 2 4 6 8 10 12 4 4 4 4 4 4 4 4 4 4
```

2.7

Considering `x <- c("a"=1, "b"=2, "c"=3, "d"=4, "e"=5)`, show how to select the third and fifth elements of `x` by using positive integers, negative integers, a logical vector, and a character vector.

```
x <- c("a"=1, "b"=2, "c"=3, "d"=4, "e"=5)
x[c(3,5)]
```

```
## c e
## 3 5
```

```
x[-c(1,2,4)]
```

```
## c e
## 3 5
```

```
x[c(F,F,T,F,T)]
```

```
## c e
## 3 5
```

```
x[c("c", "e")]
```

```
## c e
## 3 5
```

2.8

Why are `vals[c(2, 5)]` and `vals[2, 5]` different where `vals <- outer(1:5, 1:5, FUN = "/")`? How would you select fifth and ninth elements of `vals` by the use of a matrix?

```
vals <- outer(1:5, 1:5, FUN = "/")

# Because when you subset matrix with a vector, the 2d matrix behaves
# like a vector and vals[c(2, 5)] returns the elements at indices 2
# and 5 in column-major order. vals[2, 5] returns the element at row 2, column 5.

select <- matrix(ncol=2, byrow=TRUE, c(5,1,4,2))

vals[select]

## [1] 5 2
```

2.9

Consider `df <- data.frame(a=paste("Point_", 1:20), b=rep(1:4, each = 2, len = 20), c= seq(1,40,length.out = 20), stringsAsFactors = F)`. Assign "Point_undefined" to column a of all rows of `df` where column b > 1 and column c > 21 ? What is the reason of the different result that you get if you do the same operation with `df` being created with option `stringsAsFactors = T` ?

2.10

Assume `x <- matrix(1:20, ncol=2)`. What is the difference between `x[1, , drop = T]` and `x[1, , drop = F]`? Now let `y <- as.data.frame(x)`. What is the difference between `y[,1]`, `y[[1]]` and `y[1]`

2.11

What is the difference between `x["b"] <- list(NULL)` and `x["b"] <- NULL` where `x <- list(a = c(1:5), b = c(12:15))`?

2.12

Assume you have a lookup table as `lookup <- c(a = "sun", b = "rain", c="wind", u = NA)`. How would you generate the weekly weather predictions `c("sun","sun","rain", NA, "rain", "rain", "wind")` out of this lookup table?

2.13

Now assume the weather in winter lookup table is a data frame as below and we have the predictions for the next week as stored in `weeklyCast`. How would you create "weeklyTable" by the use of `rownames` function? How would you create it by the use of `match` function? How would you order the rows of lookup table by desc column?

2.14

Consider the bigDF data frame which has 1500 columns and rows. How would you select the even numbered columns named such as “Column_2”, “Column_4”, etc.? How would you select all the columns other than column 76? How would you assign 1 to 500 randomly selected diagonal indices? How can you retrieve the row and column indices of the elements which has been assigned 1? How would you select rows where columns Column_1 or Column_2 are 1 by using the subset() function?

2.15

Assume `x <- 1:20 %% 2 == 0` and `y <- 1:20 %% 5 == 0`. What are the indices of the elements that are True for both x and y? What are the indices of the elements that are True for either x or y, or both?

3 Data Import

3.1 Flat File - Q1

A csv file has numbers as column names in the first row, i.e. IDs to randomize persons. Which parameter of read.table() needs to be adjusted to read the column names as they are in the csv?

```
tmp_tidy_table <- "1_colname, 2_colname, 3_colname
3,4,5
a,b,c"
tmp_tidy_table
```

```
## [1] "1_colname, 2_colname, 3_colname\n3,4,5\na,b,c"
```

```
read.csv(text=tmp_tidy_table)
```

```
##   X1_colname X2_colname X3_colname
## 1          3          4          5
## 2          a          b          c
```

```
# Parameter `check.names`: a logical, tests for syntactically valid variable
# names
```

```
tidy_text_df <- read.csv(text=tmp_tidy_table, check.names = FALSE)
tidy_text_df
```

```
##   1_colname 2_colname 3_colname
## 1          3          4          5
## 2          a          b          c
```

3.2 Flat File - Q2

How to read the following table to have the identical() information as in tidy_txt_df from question above?

```
tmp_messy_table <- "# This line is just useless info

1_colname,2_colname,3_colname
3,4,5
```

```
a,b,c"
```

```
# To have the identical information as in the previous table we have to check  
# which lines are comments. We can do this with `comment.char` parameter.
```

```
messy_text_df <- read.csv(text = tmp_messy_table, comment.char = '#', check.names = F)  
identical(messy_text_df, tidy_text_df)
```

```
## [1] TRUE
```

3.3 Flat File - Q3

Read the hollywood.tsv (not *.csv) file into a data.table R object. What is the problem with fread()?

```
library(data.table)  
file_holly_tab <- "extdata/hollywood.tsv"  
holly <- as.data.table(read.delim(file_holly_tab), keep_row_names=T)  
head(holly)
```

```
##           Film      Genre Lead.Studio Audience..score..  
## 1:      27 Dresses  Comedy          Fox              71  
## 2: (500) Days of Summer Comedy          Fox              81  
## 3:   A Dangerous Method  Drama Independent              89  
## 4:      A Serious Man  Drama  Universal              64  
## 5: Across the Universe Romance Independent              84  
## 6:      Beginners  Comedy Independent              80  
## Profitability Rotten.Tomatoes.. Worldwide.Gross Year  
## 1:   5.3436218           40      160.308654 2008  
## 2:   8.0960000           87       60.720000 2009  
## 3:   0.4486447           79       8.972895 2011  
## 4:   4.3828571           89      30.680000 2009  
## 5:   0.6526032           54      29.367143 2007  
## 6:   4.4718750           84      14.310000 2011
```

```
class(holly)
```

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

```
holly_data_table <- fread(file_holly_tab, skip=1)  
class(holly_data_table)
```

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

```
head(holly_data_table)
```

```
##      V1           V2      V3      V4 V5      V6 V7      V8  
## 1:  1      27 Dresses  Comedy      Fox 71 5.3436218 40 160.308654  
## 2:  2 (500) Days of Summer Comedy      Fox 81 8.0960000 87 60.720000  
## 3:  3   A Dangerous Method  Drama Independent 89 0.4486447 79 8.972895  
## 4:  4      A Serious Man  Drama  Universal 64 4.3828571 89 30.680000  
## 5:  5 Across the Universe Romance Independent 84 0.6526032 54 29.367143  
## 6:  6      Beginners  Comedy Independent 80 4.4718750 84 14.310000  
##      V9  
## 1: 2008  
## 2: 2009  
## 3: 2011
```

```
## 4: 2009
## 5: 2007
## 6: 2011

holly_cn <- c("ID", colnames(read.delim(file_holly_tab, nrow= 1)))
holly_cn

## [1] "ID"          "Film"          "Genre"
## [4] "Lead.Studio"  "Audience..score.." "Profitability"
## [7] "Rotten.Tomatoes.." "Worldwide.Gross"  "Year"

# setnames(holly, holly_cn)
# head(holly)
# Difference between them: We can keep row_names. data.frame has no row_names.
```

3.4 Flat File - Q4

Who was the oldest surviving passenger of the titanic accident (titanic.csv)? Tipp: ?subset

```
tit_df <- read.csv("extdata/titanic.csv")
head(tit_df)
```

##	pclass	survived	name	sex
## 1	1	1	Allen, Miss. Elisabeth Walton	female
## 2	1	1	Allison, Master. Hudson Trevor	male
## 3	1	0	Allison, Miss. Helen Loraine	female
## 4	1	0	Allison, Mr. Hudson Joshua Creighton	male
## 5	1	0	Allison, Mrs. Hudson J C (Bessie Waldo Daniels)	female
## 6	1	1	Anderson, Mr. Harry	male

```
##      age sibsp parch ticket      fare  cabin embarked boat body
## 1 29.00     0     0  24160 211.3375    B5      S      2   NA
## 2  0.92     1     2 113781 151.5500 C22 C26      S     11   NA
## 3  2.00     1     2 113781 151.5500 C22 C26      S      NA
## 4 30.00     1     2 113781 151.5500 C22 C26      S    135
## 5 25.00     1     2 113781 151.5500 C22 C26      S      NA
## 6 48.00     0     0  19952  26.5500   E12      S      3   NA

##      home.dest
## 1      St Louis, MO
## 2 Montreal, PQ / Chesterville, ON
## 3 Montreal, PQ / Chesterville, ON
## 4 Montreal, PQ / Chesterville, ON
## 5 Montreal, PQ / Chesterville, ON
## 6      New York, NY

survivor_name <- subset(tit_df, survived==1 & age==max(age, na.rm = T), name)
survivor_age <- subset(tit_df, survived==1 & age==max(age, na.rm = T), age)

survivor_name

##      name
## 15 Barkworth, Mr. Algernon Henry Wilson

survivor_age

##      age
## 15    80
```

3.5 Excel Questions - Q1

Read only Name, Type and Total columns for only the first 10 pokemons of the pokemon.xlsx file.

3.6 Excel Question - Q2

Which athlete won most bronze medals?

3.7 Excel Question - Q3

Are the columns Gender and Event_gender consistent?

3.8 Excel Question - Q4

Which country won most medals? Which country has the highest ratio of silver medals? Use the data in the country summary sheet starting at row 147.

3.9 Excel Questions

Which countries did participate, but without winning medals?

3.10 XML - Q1

Load the XML document plant_catalog.xml. Use XPath and DOM functions to find out all unique element names in the document.

Get all plants of zone 4 and transform the data into an R list.

3.11 XML - Q2

Read the tables HTML tables from the TUM website of dates for the winter term <https://www.tum.de/en/studies/application-and-acceptance/dates-and-deadlines/dates-and-deadlines-17/> into your workspace. When are the Christmas holidays?

3.12 JSON - Q1

Read the countries.json file. Which countries have common border with Jordan? Which country has the most neighbors?

3.13 JSON - Q2

Read this JSON file about projects funded by the world bank: world_bank.json.zip. Be aware, you might need to add syntax elements like “[” and “,” to convert the file into textbook JSON format, i.e. readable by R. What was the most expensive project?

3.14 SQL - Q1

Use the extdata/Northwind.sl3 SQLite data base and retrieve a table that lists for all customers (name of the company, name of the contact person and city) all the products (name of the product) that they ordered. How many rows does this table have? Display the first 5 rows.

4 Grammar of graphics and plotting I

4.1 Setup

```
library(ggplot2)
library(data.table)
library(magrittr)
library(tidyr)

##
## Attaching package: 'tidyr'
## The following object is masked from 'package:magrittr':
##
##      extract
```

4.2 Q1

Match each chart type with the relationship it shows best.

1. shows distribution and quantiles, especially useful when comparing distributions.
2. highlights individual values, supports comparison and can show rankings or deviations categories and totals
3. shows overall changes and patterns, usually over intervals of time
4. shows relationship between two continues variables.

Options: bar chart, line chart, scatterplot, boxplot

```
# Answer
#
# 1 -> boxplot
# 2 -> bar chart
# 3 -> line chart
# 4. -> scatterplot
```

4.3 Q2

Iris is a classical dataset in machine learning literature, was first introduced by R.A. Fisher in his 1936 paper. Load the iris data into your R environment. What is the dimension of the dataset and what kind of data type does each column has?

```
dim(iris)

## [1] 150 5

head(iris)
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1          5.1         3.5         1.4         0.2  setosa
## 2          4.9         3.0         1.4         0.2  setosa
## 3          4.7         3.2         1.3         0.2  setosa
## 4          4.6         3.1         1.5         0.2  setosa
## 5          5.0         3.6         1.4         0.2  setosa
## 6          5.4         3.9         1.7         0.4  setosa
```

4.4 Q3

How are the lengths and widths of sepals and petals distributed? How would you visualize them. Describe what you see. Hint: `facet_wrap(~variable)`.

```
iris_melt <- melt(iris, id.var=c("Species"))
iris_melt
```

```
##      Species      variable value
## 1    setosa Sepal.Length    5.1
## 2    setosa Sepal.Length    4.9
## 3    setosa Sepal.Length    4.7
## 4    setosa Sepal.Length    4.6
## 5    setosa Sepal.Length    5.0
## 6    setosa Sepal.Length    5.4
## 7    setosa Sepal.Length    4.6
## 8    setosa Sepal.Length    5.0
## 9    setosa Sepal.Length    4.4
## 10   setosa Sepal.Length    4.9
## 11   setosa Sepal.Length    5.4
## 12   setosa Sepal.Length    4.8
## 13   setosa Sepal.Length    4.8
## 14   setosa Sepal.Length    4.3
## 15   setosa Sepal.Length    5.8
## 16   setosa Sepal.Length    5.7
## 17   setosa Sepal.Length    5.4
## 18   setosa Sepal.Length    5.1
## 19   setosa Sepal.Length    5.7
## 20   setosa Sepal.Length    5.1
## 21   setosa Sepal.Length    5.4
## 22   setosa Sepal.Length    5.1
## 23   setosa Sepal.Length    4.6
## 24   setosa Sepal.Length    5.1
## 25   setosa Sepal.Length    4.8
## 26   setosa Sepal.Length    5.0
## 27   setosa Sepal.Length    5.0
## 28   setosa Sepal.Length    5.2
## 29   setosa Sepal.Length    5.2
## 30   setosa Sepal.Length    4.7
## 31   setosa Sepal.Length    4.8
## 32   setosa Sepal.Length    5.4
## 33   setosa Sepal.Length    5.2
## 34   setosa Sepal.Length    5.5
## 35   setosa Sepal.Length    4.9
## 36   setosa Sepal.Length    5.0
## 37   setosa Sepal.Length    5.5
```

## 38	setosa	Sepal.Length	4.9
## 39	setosa	Sepal.Length	4.4
## 40	setosa	Sepal.Length	5.1
## 41	setosa	Sepal.Length	5.0
## 42	setosa	Sepal.Length	4.5
## 43	setosa	Sepal.Length	4.4
## 44	setosa	Sepal.Length	5.0
## 45	setosa	Sepal.Length	5.1
## 46	setosa	Sepal.Length	4.8
## 47	setosa	Sepal.Length	5.1
## 48	setosa	Sepal.Length	4.6
## 49	setosa	Sepal.Length	5.3
## 50	setosa	Sepal.Length	5.0
## 51	versicolor	Sepal.Length	7.0
## 52	versicolor	Sepal.Length	6.4
## 53	versicolor	Sepal.Length	6.9
## 54	versicolor	Sepal.Length	5.5
## 55	versicolor	Sepal.Length	6.5
## 56	versicolor	Sepal.Length	5.7
## 57	versicolor	Sepal.Length	6.3
## 58	versicolor	Sepal.Length	4.9
## 59	versicolor	Sepal.Length	6.6
## 60	versicolor	Sepal.Length	5.2
## 61	versicolor	Sepal.Length	5.0
## 62	versicolor	Sepal.Length	5.9
## 63	versicolor	Sepal.Length	6.0
## 64	versicolor	Sepal.Length	6.1
## 65	versicolor	Sepal.Length	5.6
## 66	versicolor	Sepal.Length	6.7
## 67	versicolor	Sepal.Length	5.6
## 68	versicolor	Sepal.Length	5.8
## 69	versicolor	Sepal.Length	6.2
## 70	versicolor	Sepal.Length	5.6
## 71	versicolor	Sepal.Length	5.9
## 72	versicolor	Sepal.Length	6.1
## 73	versicolor	Sepal.Length	6.3
## 74	versicolor	Sepal.Length	6.1
## 75	versicolor	Sepal.Length	6.4
## 76	versicolor	Sepal.Length	6.6
## 77	versicolor	Sepal.Length	6.8
## 78	versicolor	Sepal.Length	6.7
## 79	versicolor	Sepal.Length	6.0
## 80	versicolor	Sepal.Length	5.7
## 81	versicolor	Sepal.Length	5.5
## 82	versicolor	Sepal.Length	5.5
## 83	versicolor	Sepal.Length	5.8
## 84	versicolor	Sepal.Length	6.0
## 85	versicolor	Sepal.Length	5.4
## 86	versicolor	Sepal.Length	6.0
## 87	versicolor	Sepal.Length	6.7
## 88	versicolor	Sepal.Length	6.3
## 89	versicolor	Sepal.Length	5.6
## 90	versicolor	Sepal.Length	5.5
## 91	versicolor	Sepal.Length	5.5

## 92	versicolor	Sepal.Length	6.1
## 93	versicolor	Sepal.Length	5.8
## 94	versicolor	Sepal.Length	5.0
## 95	versicolor	Sepal.Length	5.6
## 96	versicolor	Sepal.Length	5.7
## 97	versicolor	Sepal.Length	5.7
## 98	versicolor	Sepal.Length	6.2
## 99	versicolor	Sepal.Length	5.1
## 100	versicolor	Sepal.Length	5.7
## 101	virginica	Sepal.Length	6.3
## 102	virginica	Sepal.Length	5.8
## 103	virginica	Sepal.Length	7.1
## 104	virginica	Sepal.Length	6.3
## 105	virginica	Sepal.Length	6.5
## 106	virginica	Sepal.Length	7.6
## 107	virginica	Sepal.Length	4.9
## 108	virginica	Sepal.Length	7.3
## 109	virginica	Sepal.Length	6.7
## 110	virginica	Sepal.Length	7.2
## 111	virginica	Sepal.Length	6.5
## 112	virginica	Sepal.Length	6.4
## 113	virginica	Sepal.Length	6.8
## 114	virginica	Sepal.Length	5.7
## 115	virginica	Sepal.Length	5.8
## 116	virginica	Sepal.Length	6.4
## 117	virginica	Sepal.Length	6.5
## 118	virginica	Sepal.Length	7.7
## 119	virginica	Sepal.Length	7.7
## 120	virginica	Sepal.Length	6.0
## 121	virginica	Sepal.Length	6.9
## 122	virginica	Sepal.Length	5.6
## 123	virginica	Sepal.Length	7.7
## 124	virginica	Sepal.Length	6.3
## 125	virginica	Sepal.Length	6.7
## 126	virginica	Sepal.Length	7.2
## 127	virginica	Sepal.Length	6.2
## 128	virginica	Sepal.Length	6.1
## 129	virginica	Sepal.Length	6.4
## 130	virginica	Sepal.Length	7.2
## 131	virginica	Sepal.Length	7.4
## 132	virginica	Sepal.Length	7.9
## 133	virginica	Sepal.Length	6.4
## 134	virginica	Sepal.Length	6.3
## 135	virginica	Sepal.Length	6.1
## 136	virginica	Sepal.Length	7.7
## 137	virginica	Sepal.Length	6.3
## 138	virginica	Sepal.Length	6.4
## 139	virginica	Sepal.Length	6.0
## 140	virginica	Sepal.Length	6.9
## 141	virginica	Sepal.Length	6.7
## 142	virginica	Sepal.Length	6.9
## 143	virginica	Sepal.Length	5.8
## 144	virginica	Sepal.Length	6.8
## 145	virginica	Sepal.Length	6.7

## 146	virginica	Sepal.Length	6.7
## 147	virginica	Sepal.Length	6.3
## 148	virginica	Sepal.Length	6.5
## 149	virginica	Sepal.Length	6.2
## 150	virginica	Sepal.Length	5.9
## 151	setosa	Sepal.Width	3.5
## 152	setosa	Sepal.Width	3.0
## 153	setosa	Sepal.Width	3.2
## 154	setosa	Sepal.Width	3.1
## 155	setosa	Sepal.Width	3.6
## 156	setosa	Sepal.Width	3.9
## 157	setosa	Sepal.Width	3.4
## 158	setosa	Sepal.Width	3.4
## 159	setosa	Sepal.Width	2.9
## 160	setosa	Sepal.Width	3.1
## 161	setosa	Sepal.Width	3.7
## 162	setosa	Sepal.Width	3.4
## 163	setosa	Sepal.Width	3.0
## 164	setosa	Sepal.Width	3.0
## 165	setosa	Sepal.Width	4.0
## 166	setosa	Sepal.Width	4.4
## 167	setosa	Sepal.Width	3.9
## 168	setosa	Sepal.Width	3.5
## 169	setosa	Sepal.Width	3.8
## 170	setosa	Sepal.Width	3.8
## 171	setosa	Sepal.Width	3.4
## 172	setosa	Sepal.Width	3.7
## 173	setosa	Sepal.Width	3.6
## 174	setosa	Sepal.Width	3.3
## 175	setosa	Sepal.Width	3.4
## 176	setosa	Sepal.Width	3.0
## 177	setosa	Sepal.Width	3.4
## 178	setosa	Sepal.Width	3.5
## 179	setosa	Sepal.Width	3.4
## 180	setosa	Sepal.Width	3.2
## 181	setosa	Sepal.Width	3.1
## 182	setosa	Sepal.Width	3.4
## 183	setosa	Sepal.Width	4.1
## 184	setosa	Sepal.Width	4.2
## 185	setosa	Sepal.Width	3.1
## 186	setosa	Sepal.Width	3.2
## 187	setosa	Sepal.Width	3.5
## 188	setosa	Sepal.Width	3.6
## 189	setosa	Sepal.Width	3.0
## 190	setosa	Sepal.Width	3.4
## 191	setosa	Sepal.Width	3.5
## 192	setosa	Sepal.Width	2.3
## 193	setosa	Sepal.Width	3.2
## 194	setosa	Sepal.Width	3.5
## 195	setosa	Sepal.Width	3.8
## 196	setosa	Sepal.Width	3.0
## 197	setosa	Sepal.Width	3.8
## 198	setosa	Sepal.Width	3.2
## 199	setosa	Sepal.Width	3.7

##	200	setosa	Sepal.Width	3.3
##	201	versicolor	Sepal.Width	3.2
##	202	versicolor	Sepal.Width	3.2
##	203	versicolor	Sepal.Width	3.1
##	204	versicolor	Sepal.Width	2.3
##	205	versicolor	Sepal.Width	2.8
##	206	versicolor	Sepal.Width	2.8
##	207	versicolor	Sepal.Width	3.3
##	208	versicolor	Sepal.Width	2.4
##	209	versicolor	Sepal.Width	2.9
##	210	versicolor	Sepal.Width	2.7
##	211	versicolor	Sepal.Width	2.0
##	212	versicolor	Sepal.Width	3.0
##	213	versicolor	Sepal.Width	2.2
##	214	versicolor	Sepal.Width	2.9
##	215	versicolor	Sepal.Width	2.9
##	216	versicolor	Sepal.Width	3.1
##	217	versicolor	Sepal.Width	3.0
##	218	versicolor	Sepal.Width	2.7
##	219	versicolor	Sepal.Width	2.2
##	220	versicolor	Sepal.Width	2.5
##	221	versicolor	Sepal.Width	3.2
##	222	versicolor	Sepal.Width	2.8
##	223	versicolor	Sepal.Width	2.5
##	224	versicolor	Sepal.Width	2.8
##	225	versicolor	Sepal.Width	2.9
##	226	versicolor	Sepal.Width	3.0
##	227	versicolor	Sepal.Width	2.8
##	228	versicolor	Sepal.Width	3.0
##	229	versicolor	Sepal.Width	2.9
##	230	versicolor	Sepal.Width	2.6
##	231	versicolor	Sepal.Width	2.4
##	232	versicolor	Sepal.Width	2.4
##	233	versicolor	Sepal.Width	2.7
##	234	versicolor	Sepal.Width	2.7
##	235	versicolor	Sepal.Width	3.0
##	236	versicolor	Sepal.Width	3.4
##	237	versicolor	Sepal.Width	3.1
##	238	versicolor	Sepal.Width	2.3
##	239	versicolor	Sepal.Width	3.0
##	240	versicolor	Sepal.Width	2.5
##	241	versicolor	Sepal.Width	2.6
##	242	versicolor	Sepal.Width	3.0
##	243	versicolor	Sepal.Width	2.6
##	244	versicolor	Sepal.Width	2.3
##	245	versicolor	Sepal.Width	2.7
##	246	versicolor	Sepal.Width	3.0
##	247	versicolor	Sepal.Width	2.9
##	248	versicolor	Sepal.Width	2.9
##	249	versicolor	Sepal.Width	2.5
##	250	versicolor	Sepal.Width	2.8
##	251	virginica	Sepal.Width	3.3
##	252	virginica	Sepal.Width	2.7
##	253	virginica	Sepal.Width	3.0

##	254	virginica	Sepal.Width	2.9
##	255	virginica	Sepal.Width	3.0
##	256	virginica	Sepal.Width	3.0
##	257	virginica	Sepal.Width	2.5
##	258	virginica	Sepal.Width	2.9
##	259	virginica	Sepal.Width	2.5
##	260	virginica	Sepal.Width	3.6
##	261	virginica	Sepal.Width	3.2
##	262	virginica	Sepal.Width	2.7
##	263	virginica	Sepal.Width	3.0
##	264	virginica	Sepal.Width	2.5
##	265	virginica	Sepal.Width	2.8
##	266	virginica	Sepal.Width	3.2
##	267	virginica	Sepal.Width	3.0
##	268	virginica	Sepal.Width	3.8
##	269	virginica	Sepal.Width	2.6
##	270	virginica	Sepal.Width	2.2
##	271	virginica	Sepal.Width	3.2
##	272	virginica	Sepal.Width	2.8
##	273	virginica	Sepal.Width	2.8
##	274	virginica	Sepal.Width	2.7
##	275	virginica	Sepal.Width	3.3
##	276	virginica	Sepal.Width	3.2
##	277	virginica	Sepal.Width	2.8
##	278	virginica	Sepal.Width	3.0
##	279	virginica	Sepal.Width	2.8
##	280	virginica	Sepal.Width	3.0
##	281	virginica	Sepal.Width	2.8
##	282	virginica	Sepal.Width	3.8
##	283	virginica	Sepal.Width	2.8
##	284	virginica	Sepal.Width	2.8
##	285	virginica	Sepal.Width	2.6
##	286	virginica	Sepal.Width	3.0
##	287	virginica	Sepal.Width	3.4
##	288	virginica	Sepal.Width	3.1
##	289	virginica	Sepal.Width	3.0
##	290	virginica	Sepal.Width	3.1
##	291	virginica	Sepal.Width	3.1
##	292	virginica	Sepal.Width	3.1
##	293	virginica	Sepal.Width	2.7
##	294	virginica	Sepal.Width	3.2
##	295	virginica	Sepal.Width	3.3
##	296	virginica	Sepal.Width	3.0
##	297	virginica	Sepal.Width	2.5
##	298	virginica	Sepal.Width	3.0
##	299	virginica	Sepal.Width	3.4
##	300	virginica	Sepal.Width	3.0
##	301	setosa	Petal.Length	1.4
##	302	setosa	Petal.Length	1.4
##	303	setosa	Petal.Length	1.3
##	304	setosa	Petal.Length	1.5
##	305	setosa	Petal.Length	1.4
##	306	setosa	Petal.Length	1.7
##	307	setosa	Petal.Length	1.4

## 308	setosa	Petal.Length	1.5
## 309	setosa	Petal.Length	1.4
## 310	setosa	Petal.Length	1.5
## 311	setosa	Petal.Length	1.5
## 312	setosa	Petal.Length	1.6
## 313	setosa	Petal.Length	1.4
## 314	setosa	Petal.Length	1.1
## 315	setosa	Petal.Length	1.2
## 316	setosa	Petal.Length	1.5
## 317	setosa	Petal.Length	1.3
## 318	setosa	Petal.Length	1.4
## 319	setosa	Petal.Length	1.7
## 320	setosa	Petal.Length	1.5
## 321	setosa	Petal.Length	1.7
## 322	setosa	Petal.Length	1.5
## 323	setosa	Petal.Length	1.0
## 324	setosa	Petal.Length	1.7
## 325	setosa	Petal.Length	1.9
## 326	setosa	Petal.Length	1.6
## 327	setosa	Petal.Length	1.6
## 328	setosa	Petal.Length	1.5
## 329	setosa	Petal.Length	1.4
## 330	setosa	Petal.Length	1.6
## 331	setosa	Petal.Length	1.6
## 332	setosa	Petal.Length	1.5
## 333	setosa	Petal.Length	1.5
## 334	setosa	Petal.Length	1.4
## 335	setosa	Petal.Length	1.5
## 336	setosa	Petal.Length	1.2
## 337	setosa	Petal.Length	1.3
## 338	setosa	Petal.Length	1.4
## 339	setosa	Petal.Length	1.3
## 340	setosa	Petal.Length	1.5
## 341	setosa	Petal.Length	1.3
## 342	setosa	Petal.Length	1.3
## 343	setosa	Petal.Length	1.3
## 344	setosa	Petal.Length	1.6
## 345	setosa	Petal.Length	1.9
## 346	setosa	Petal.Length	1.4
## 347	setosa	Petal.Length	1.6
## 348	setosa	Petal.Length	1.4
## 349	setosa	Petal.Length	1.5
## 350	setosa	Petal.Length	1.4
## 351	versicolor	Petal.Length	4.7
## 352	versicolor	Petal.Length	4.5
## 353	versicolor	Petal.Length	4.9
## 354	versicolor	Petal.Length	4.0
## 355	versicolor	Petal.Length	4.6
## 356	versicolor	Petal.Length	4.5
## 357	versicolor	Petal.Length	4.7
## 358	versicolor	Petal.Length	3.3
## 359	versicolor	Petal.Length	4.6
## 360	versicolor	Petal.Length	3.9
## 361	versicolor	Petal.Length	3.5

## 362	versicolor	Petal.Length	4.2
## 363	versicolor	Petal.Length	4.0
## 364	versicolor	Petal.Length	4.7
## 365	versicolor	Petal.Length	3.6
## 366	versicolor	Petal.Length	4.4
## 367	versicolor	Petal.Length	4.5
## 368	versicolor	Petal.Length	4.1
## 369	versicolor	Petal.Length	4.5
## 370	versicolor	Petal.Length	3.9
## 371	versicolor	Petal.Length	4.8
## 372	versicolor	Petal.Length	4.0
## 373	versicolor	Petal.Length	4.9
## 374	versicolor	Petal.Length	4.7
## 375	versicolor	Petal.Length	4.3
## 376	versicolor	Petal.Length	4.4
## 377	versicolor	Petal.Length	4.8
## 378	versicolor	Petal.Length	5.0
## 379	versicolor	Petal.Length	4.5
## 380	versicolor	Petal.Length	3.5
## 381	versicolor	Petal.Length	3.8
## 382	versicolor	Petal.Length	3.7
## 383	versicolor	Petal.Length	3.9
## 384	versicolor	Petal.Length	5.1
## 385	versicolor	Petal.Length	4.5
## 386	versicolor	Petal.Length	4.5
## 387	versicolor	Petal.Length	4.7
## 388	versicolor	Petal.Length	4.4
## 389	versicolor	Petal.Length	4.1
## 390	versicolor	Petal.Length	4.0
## 391	versicolor	Petal.Length	4.4
## 392	versicolor	Petal.Length	4.6
## 393	versicolor	Petal.Length	4.0
## 394	versicolor	Petal.Length	3.3
## 395	versicolor	Petal.Length	4.2
## 396	versicolor	Petal.Length	4.2
## 397	versicolor	Petal.Length	4.2
## 398	versicolor	Petal.Length	4.3
## 399	versicolor	Petal.Length	3.0
## 400	versicolor	Petal.Length	4.1
## 401	virginica	Petal.Length	6.0
## 402	virginica	Petal.Length	5.1
## 403	virginica	Petal.Length	5.9
## 404	virginica	Petal.Length	5.6
## 405	virginica	Petal.Length	5.8
## 406	virginica	Petal.Length	6.6
## 407	virginica	Petal.Length	4.5
## 408	virginica	Petal.Length	6.3
## 409	virginica	Petal.Length	5.8
## 410	virginica	Petal.Length	6.1
## 411	virginica	Petal.Length	5.1
## 412	virginica	Petal.Length	5.3
## 413	virginica	Petal.Length	5.5
## 414	virginica	Petal.Length	5.0
## 415	virginica	Petal.Length	5.1

## 416	virginica	Petal.Length	5.3
## 417	virginica	Petal.Length	5.5
## 418	virginica	Petal.Length	6.7
## 419	virginica	Petal.Length	6.9
## 420	virginica	Petal.Length	5.0
## 421	virginica	Petal.Length	5.7
## 422	virginica	Petal.Length	4.9
## 423	virginica	Petal.Length	6.7
## 424	virginica	Petal.Length	4.9
## 425	virginica	Petal.Length	5.7
## 426	virginica	Petal.Length	6.0
## 427	virginica	Petal.Length	4.8
## 428	virginica	Petal.Length	4.9
## 429	virginica	Petal.Length	5.6
## 430	virginica	Petal.Length	5.8
## 431	virginica	Petal.Length	6.1
## 432	virginica	Petal.Length	6.4
## 433	virginica	Petal.Length	5.6
## 434	virginica	Petal.Length	5.1
## 435	virginica	Petal.Length	5.6
## 436	virginica	Petal.Length	6.1
## 437	virginica	Petal.Length	5.6
## 438	virginica	Petal.Length	5.5
## 439	virginica	Petal.Length	4.8
## 440	virginica	Petal.Length	5.4
## 441	virginica	Petal.Length	5.6
## 442	virginica	Petal.Length	5.1
## 443	virginica	Petal.Length	5.1
## 444	virginica	Petal.Length	5.9
## 445	virginica	Petal.Length	5.7
## 446	virginica	Petal.Length	5.2
## 447	virginica	Petal.Length	5.0
## 448	virginica	Petal.Length	5.2
## 449	virginica	Petal.Length	5.4
## 450	virginica	Petal.Length	5.1
## 451	setosa	Petal.Width	0.2
## 452	setosa	Petal.Width	0.2
## 453	setosa	Petal.Width	0.2
## 454	setosa	Petal.Width	0.2
## 455	setosa	Petal.Width	0.2
## 456	setosa	Petal.Width	0.4
## 457	setosa	Petal.Width	0.3
## 458	setosa	Petal.Width	0.2
## 459	setosa	Petal.Width	0.2
## 460	setosa	Petal.Width	0.1
## 461	setosa	Petal.Width	0.2
## 462	setosa	Petal.Width	0.2
## 463	setosa	Petal.Width	0.1
## 464	setosa	Petal.Width	0.1
## 465	setosa	Petal.Width	0.2
## 466	setosa	Petal.Width	0.4
## 467	setosa	Petal.Width	0.4
## 468	setosa	Petal.Width	0.3
## 469	setosa	Petal.Width	0.3

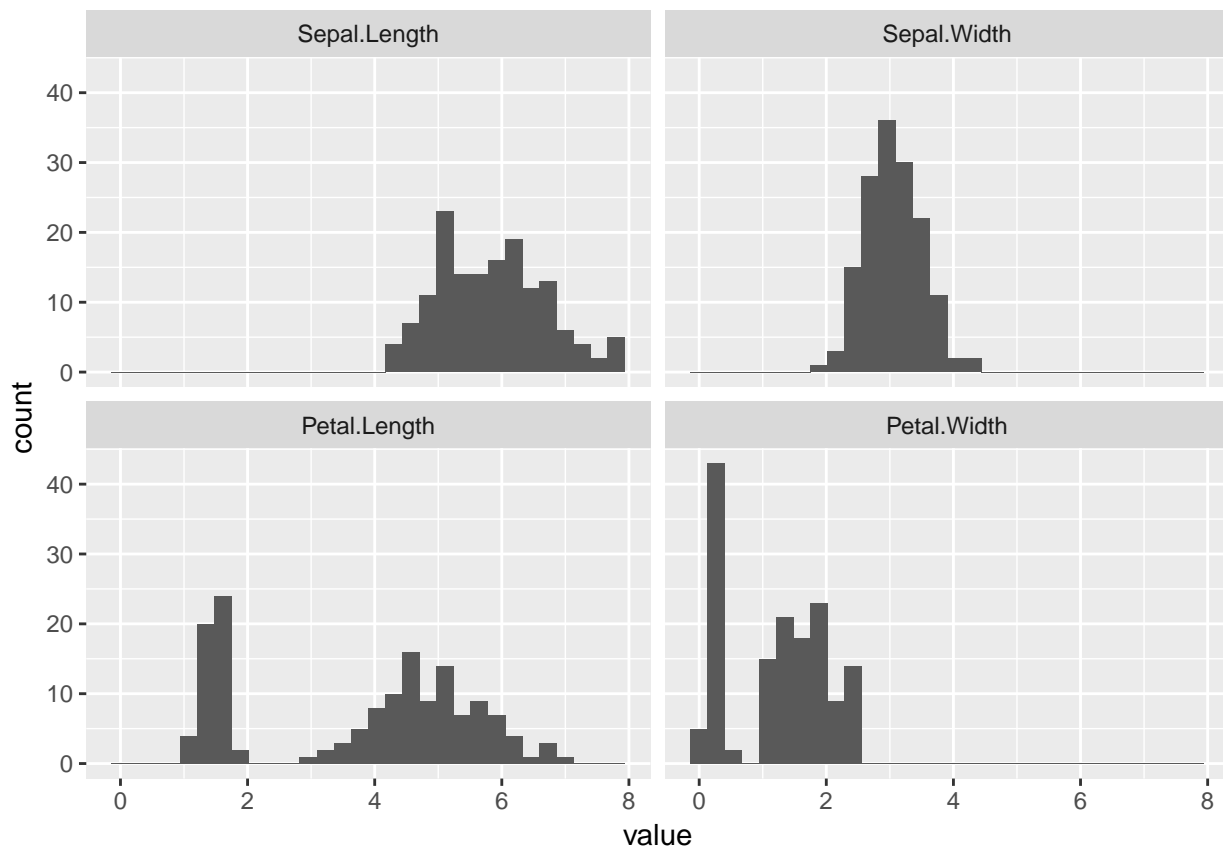
## 470	setosa	Petal.Width	0.3
## 471	setosa	Petal.Width	0.2
## 472	setosa	Petal.Width	0.4
## 473	setosa	Petal.Width	0.2
## 474	setosa	Petal.Width	0.5
## 475	setosa	Petal.Width	0.2
## 476	setosa	Petal.Width	0.2
## 477	setosa	Petal.Width	0.4
## 478	setosa	Petal.Width	0.2
## 479	setosa	Petal.Width	0.2
## 480	setosa	Petal.Width	0.2
## 481	setosa	Petal.Width	0.2
## 482	setosa	Petal.Width	0.4
## 483	setosa	Petal.Width	0.1
## 484	setosa	Petal.Width	0.2
## 485	setosa	Petal.Width	0.2
## 486	setosa	Petal.Width	0.2
## 487	setosa	Petal.Width	0.2
## 488	setosa	Petal.Width	0.1
## 489	setosa	Petal.Width	0.2
## 490	setosa	Petal.Width	0.2
## 491	setosa	Petal.Width	0.3
## 492	setosa	Petal.Width	0.3
## 493	setosa	Petal.Width	0.2
## 494	setosa	Petal.Width	0.6
## 495	setosa	Petal.Width	0.4
## 496	setosa	Petal.Width	0.3
## 497	setosa	Petal.Width	0.2
## 498	setosa	Petal.Width	0.2
## 499	setosa	Petal.Width	0.2
## 500	setosa	Petal.Width	0.2
## 501	versicolor	Petal.Width	1.4
## 502	versicolor	Petal.Width	1.5
## 503	versicolor	Petal.Width	1.5
## 504	versicolor	Petal.Width	1.3
## 505	versicolor	Petal.Width	1.5
## 506	versicolor	Petal.Width	1.3
## 507	versicolor	Petal.Width	1.6
## 508	versicolor	Petal.Width	1.0
## 509	versicolor	Petal.Width	1.3
## 510	versicolor	Petal.Width	1.4
## 511	versicolor	Petal.Width	1.0
## 512	versicolor	Petal.Width	1.5
## 513	versicolor	Petal.Width	1.0
## 514	versicolor	Petal.Width	1.4
## 515	versicolor	Petal.Width	1.3
## 516	versicolor	Petal.Width	1.4
## 517	versicolor	Petal.Width	1.5
## 518	versicolor	Petal.Width	1.0
## 519	versicolor	Petal.Width	1.5
## 520	versicolor	Petal.Width	1.1
## 521	versicolor	Petal.Width	1.8
## 522	versicolor	Petal.Width	1.3
## 523	versicolor	Petal.Width	1.5

##	524	versicolor	Petal.Width	1.2
##	525	versicolor	Petal.Width	1.3
##	526	versicolor	Petal.Width	1.4
##	527	versicolor	Petal.Width	1.4
##	528	versicolor	Petal.Width	1.7
##	529	versicolor	Petal.Width	1.5
##	530	versicolor	Petal.Width	1.0
##	531	versicolor	Petal.Width	1.1
##	532	versicolor	Petal.Width	1.0
##	533	versicolor	Petal.Width	1.2
##	534	versicolor	Petal.Width	1.6
##	535	versicolor	Petal.Width	1.5
##	536	versicolor	Petal.Width	1.6
##	537	versicolor	Petal.Width	1.5
##	538	versicolor	Petal.Width	1.3
##	539	versicolor	Petal.Width	1.3
##	540	versicolor	Petal.Width	1.3
##	541	versicolor	Petal.Width	1.2
##	542	versicolor	Petal.Width	1.4
##	543	versicolor	Petal.Width	1.2
##	544	versicolor	Petal.Width	1.0
##	545	versicolor	Petal.Width	1.3
##	546	versicolor	Petal.Width	1.2
##	547	versicolor	Petal.Width	1.3
##	548	versicolor	Petal.Width	1.3
##	549	versicolor	Petal.Width	1.1
##	550	versicolor	Petal.Width	1.3
##	551	virginica	Petal.Width	2.5
##	552	virginica	Petal.Width	1.9
##	553	virginica	Petal.Width	2.1
##	554	virginica	Petal.Width	1.8
##	555	virginica	Petal.Width	2.2
##	556	virginica	Petal.Width	2.1
##	557	virginica	Petal.Width	1.7
##	558	virginica	Petal.Width	1.8
##	559	virginica	Petal.Width	1.8
##	560	virginica	Petal.Width	2.5
##	561	virginica	Petal.Width	2.0
##	562	virginica	Petal.Width	1.9
##	563	virginica	Petal.Width	2.1
##	564	virginica	Petal.Width	2.0
##	565	virginica	Petal.Width	2.4
##	566	virginica	Petal.Width	2.3
##	567	virginica	Petal.Width	1.8
##	568	virginica	Petal.Width	2.2
##	569	virginica	Petal.Width	2.3
##	570	virginica	Petal.Width	1.5
##	571	virginica	Petal.Width	2.3
##	572	virginica	Petal.Width	2.0
##	573	virginica	Petal.Width	2.0
##	574	virginica	Petal.Width	1.8
##	575	virginica	Petal.Width	2.1
##	576	virginica	Petal.Width	1.8
##	577	virginica	Petal.Width	1.8


```
## 578 virginica Petal.Width 1.8
## 579 virginica Petal.Width 2.1
## 580 virginica Petal.Width 1.6
## 581 virginica Petal.Width 1.9
## 582 virginica Petal.Width 2.0
## 583 virginica Petal.Width 2.2
## 584 virginica Petal.Width 1.5
## 585 virginica Petal.Width 1.4
## 586 virginica Petal.Width 2.3
## 587 virginica Petal.Width 2.4
## 588 virginica Petal.Width 1.8
## 589 virginica Petal.Width 1.8
## 590 virginica Petal.Width 2.1
## 591 virginica Petal.Width 2.4
## 592 virginica Petal.Width 2.3
## 593 virginica Petal.Width 1.9
## 594 virginica Petal.Width 2.3
## 595 virginica Petal.Width 2.5
## 596 virginica Petal.Width 2.3
## 597 virginica Petal.Width 1.9
## 598 virginica Petal.Width 2.0
## 599 virginica Petal.Width 2.3
## 600 virginica Petal.Width 1.8
```

```
iris_melt %>%
  ggplot(aes(value)) +
  geom_histogram() +
  facet_wrap(~variable)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

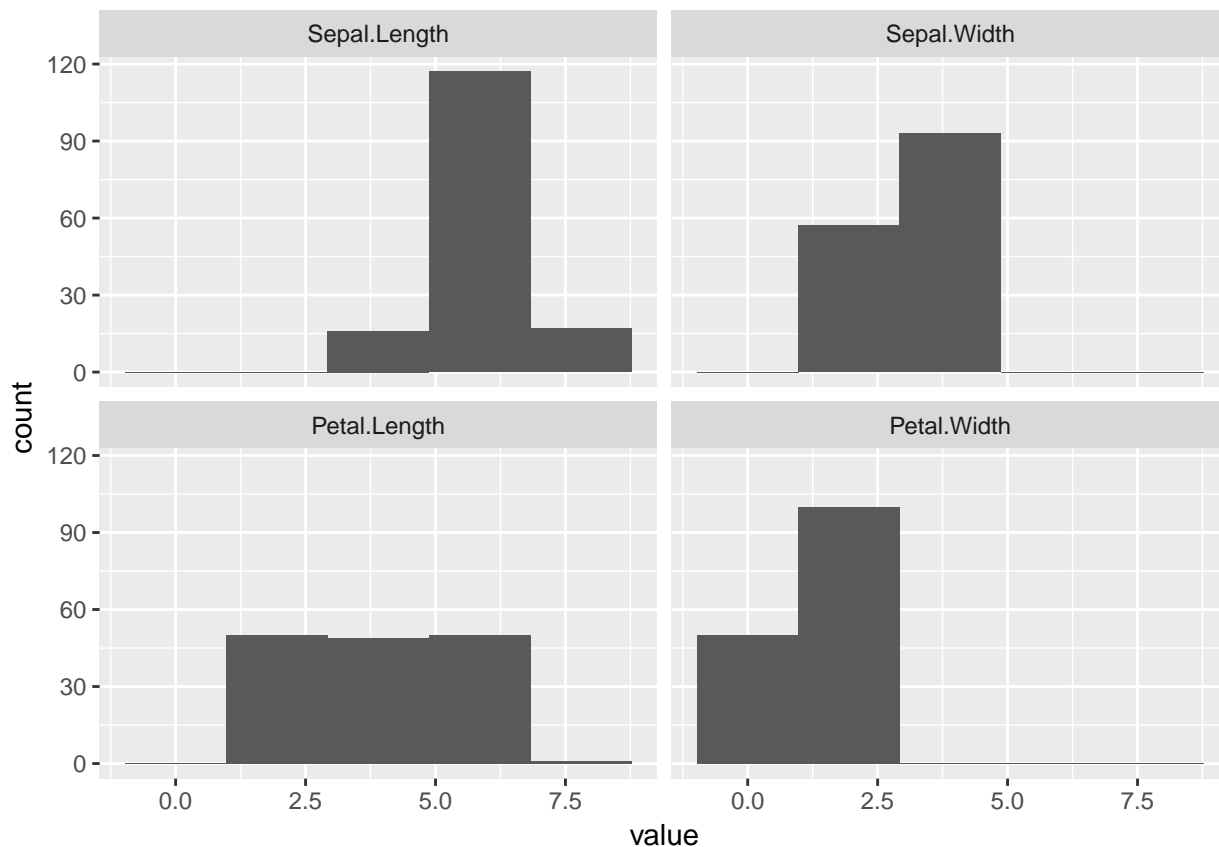


4.5 Q4

Vary the number of bins in the above histogram. Describe what you see

#Answer: With very few bins, we cannot show the bimodal distribution correctly

```
iris_melt %>%
  ggplot(aes(value)) +
  geom_histogram(bins=5) +
  facet_wrap(~variable)
```



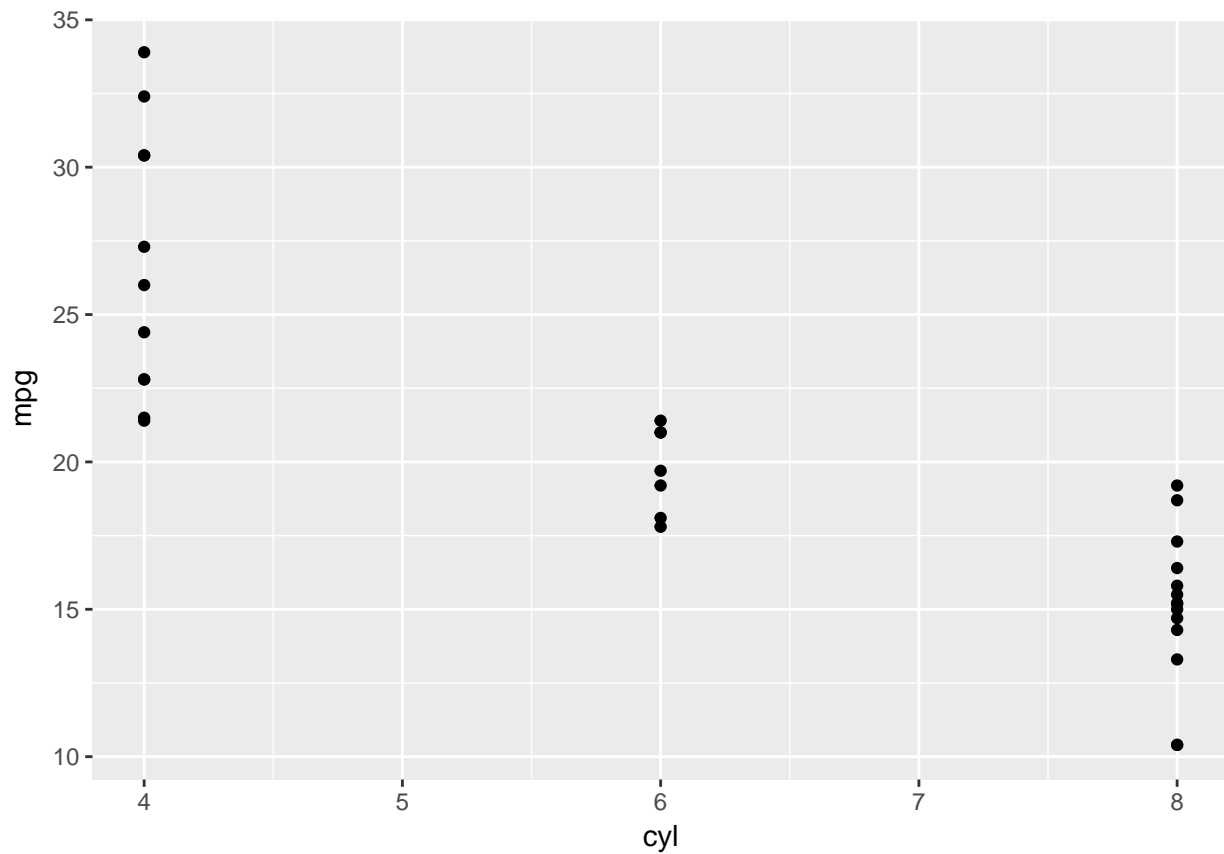
5 Datacamp

5.1 ggplot2 on Datacamp

```
library(ggplot2)
str(mtcars)
```

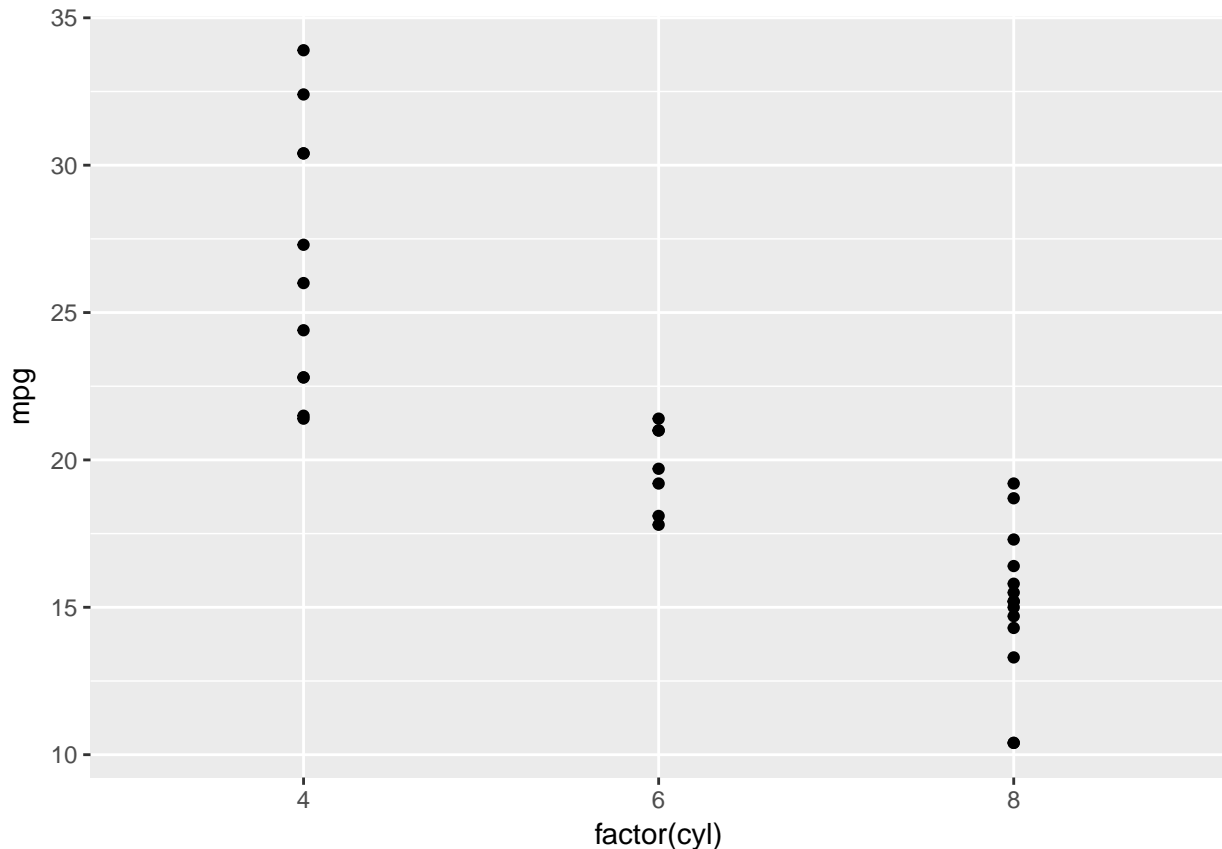
```
## 'data.frame': 32 obs. of 11 variables:
## $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : num 6 6 4 6 8 6 8 4 4 6 ...
## $ disp: num 160 160 108 258 360 ...
## $ hp : num 110 110 93 110 175 105 245 62 95 123 ...
## $ drat: num 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
## $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num 16.5 17 18.6 19.4 17 ...
## $ vs : num 0 0 1 1 0 1 0 1 1 1 ...
## $ am : num 1 1 1 0 0 0 0 0 0 0 ...
## $ gear: num 4 4 4 3 3 3 3 4 4 4 ...
## $ carb: num 4 4 1 1 2 1 4 2 2 4 ...
```

```
ggplot(mtcars, aes(x = cyl, y = mpg)) +
  geom_point()
```



*#Stellar scatterplotting! Notice that ggplot2 treats cyl as a factor.
#This time the x-axis does not contain variables like 5 or 7, only the values
#that are present in the dataset.*

```
# Change the command below so that cyl is treated as factor  
ggplot(mtcars, aes(x = factor(cyl), y = mpg)) +  
  geom_point()
```



5.2 Exploring ggplot2, part 3

We'll use several datasets throughout the courses to showcase the concepts discussed in the videos. In the previous exercises, you already got to know mtcars. Let's dive a little deeper to explore the three main topics in this course: The data, aesthetics, and geom layers.

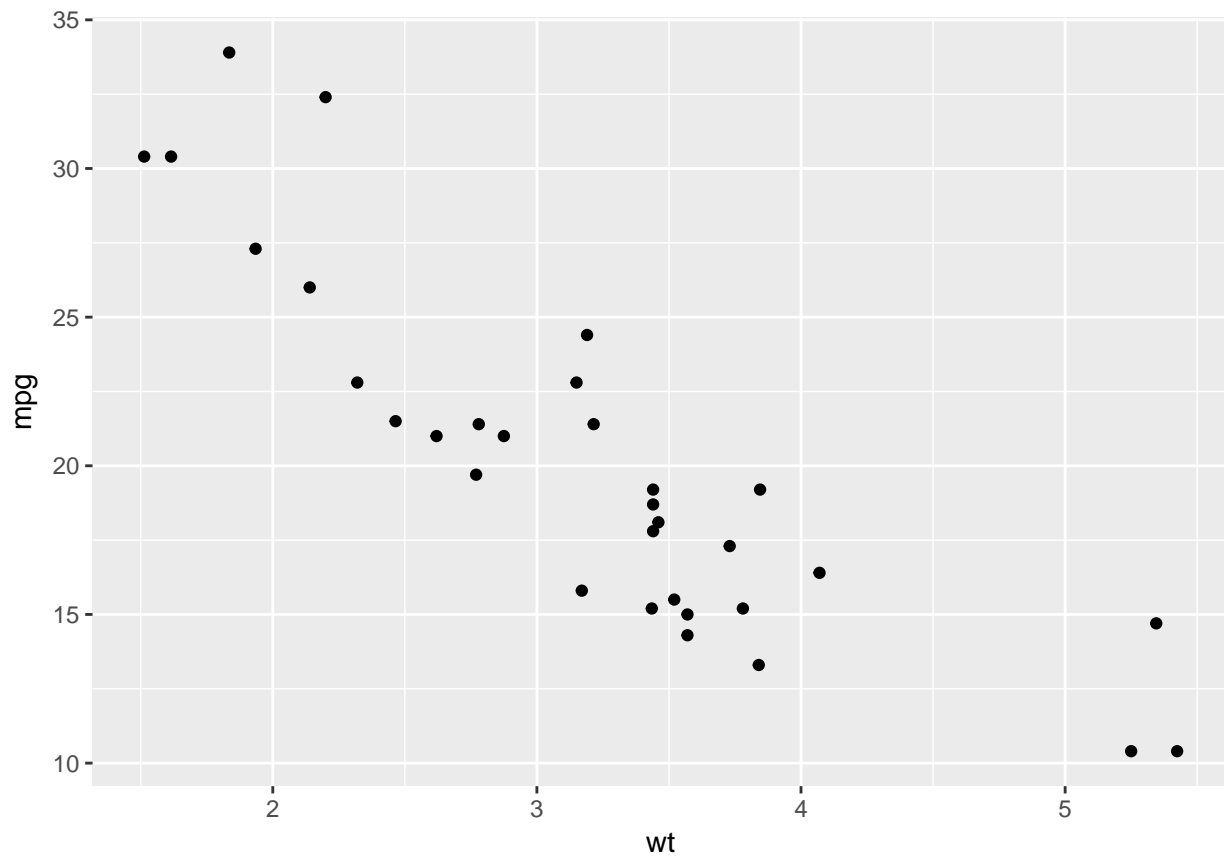
The mtcars dataset contains information about 32 cars from 1973 Motor Trend magazine. This dataset is small, intuitive, and contains a variety of continuous and categorical variables.

You're encouraged to think about how the examples and concepts we discuss throughout these data viz courses apply to your own data-sets!

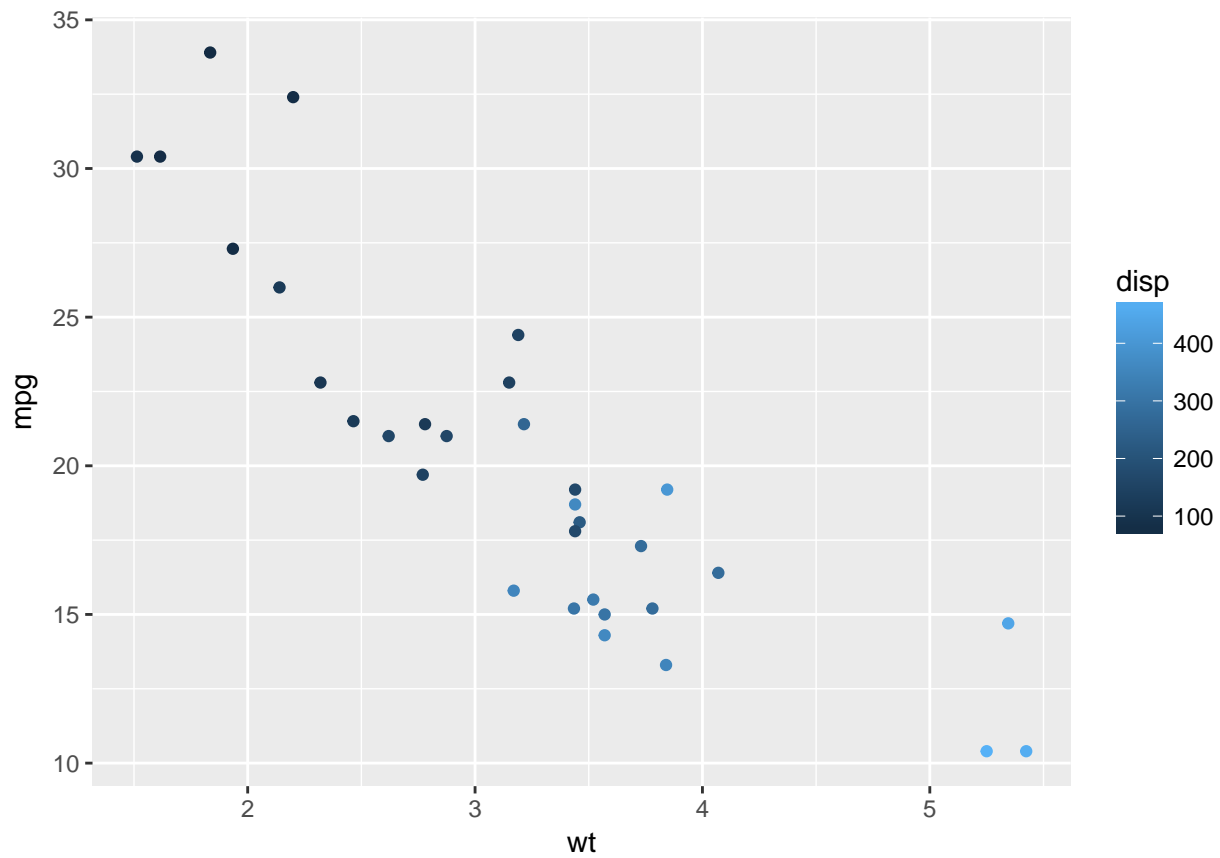
5.2.1 Instructions

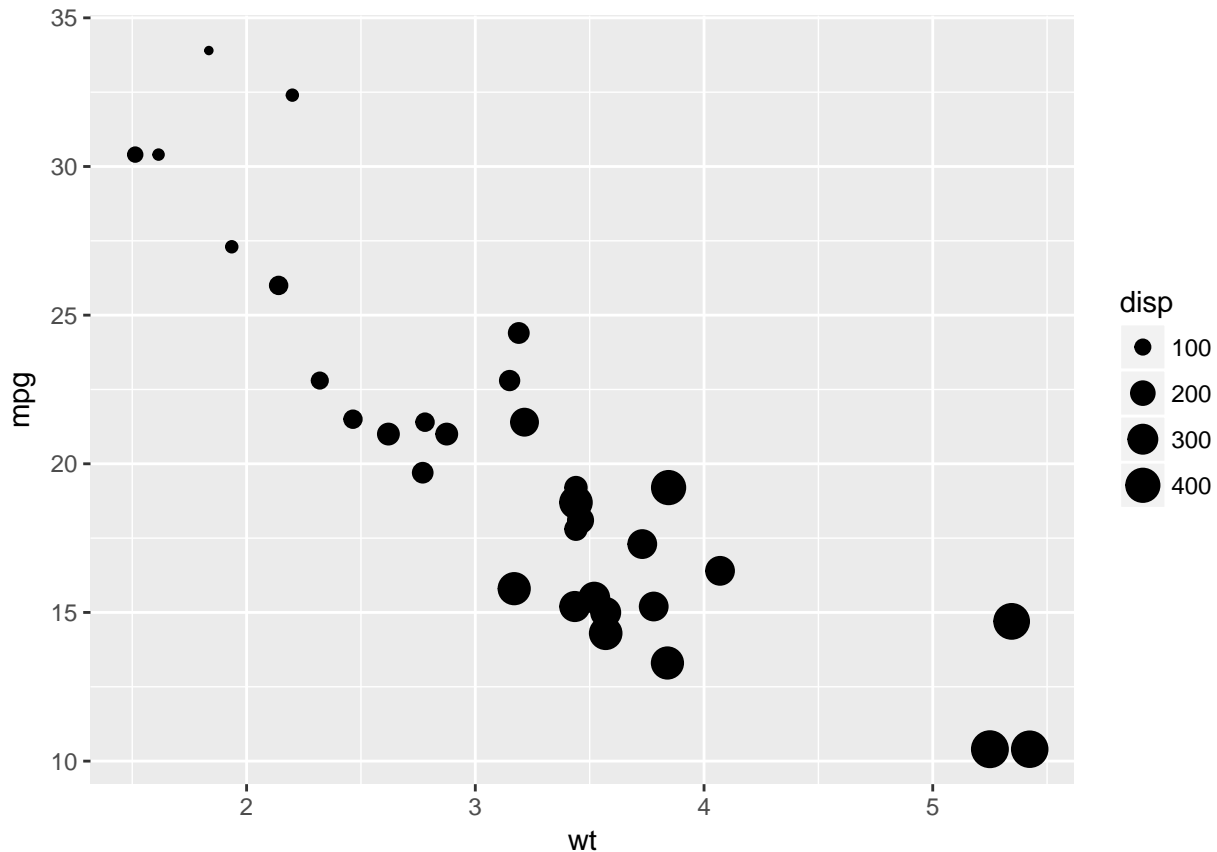
- `ggplot2` has already been loaded for you. Take a look at the first command. It plots the mpg (miles per gallon) against the weight (in thousands of pounds). You don't have to change anything about this command.
- In the second call of `ggplot()` change the color argument in `aes()` (which stands for aesthetics). The color should be dependent on the displacement of the car engine, found in `disp`.
- In the third call of `ggplot()` change the size argument in `aes()` (which stands for aesthetics). The size should be dependent on the displacement of the car engine, found in `disp`.

```
# A scatter plot has been made for you
ggplot(mtcars, aes(x = wt, y = mpg)) +
  geom_point()
```



```
# Replace ___ with the correct column  
ggplot(mtcars, aes(x = wt, y = mpg, color = disp)) +  
  geom_point()
```





5.3 Understanding Variables

In the previous exercise you saw that `disp` can be mapped onto a color gradient or onto a continuous size scale.

Another argument of `aes()` is the shape of the points. There are a finite number of shapes which `ggplot()` can automatically assign to the points. However, if you try this command in the console to the right:

```
#ggplot(mtcars, aes(x = wt, y = mpg, shape = disp)) +
#  geom_point()
```

It gives an error. What does this mean?

```
# Error: A continuous variable can not be mapped to shape
#
# Correct. The error message 'A continuous variable can not be mapped to shape',
# means that shape doesn't exist on a continuous scale here.
```

5.4 Exploring ggplot2, part 4

The `diamonds` data frame contains information on the prices and various metrics of 50,000 diamonds. Among the variables included are `carat` (a measurement of the size of the diamond) and `price`. For the next exercises, you'll be using a subset of 1,000 diamonds.

Here you'll use two common geom layer functions: `geom_point()` and `geom_smooth()`. We already saw in the earlier exercises how these are added using the `+` operator.

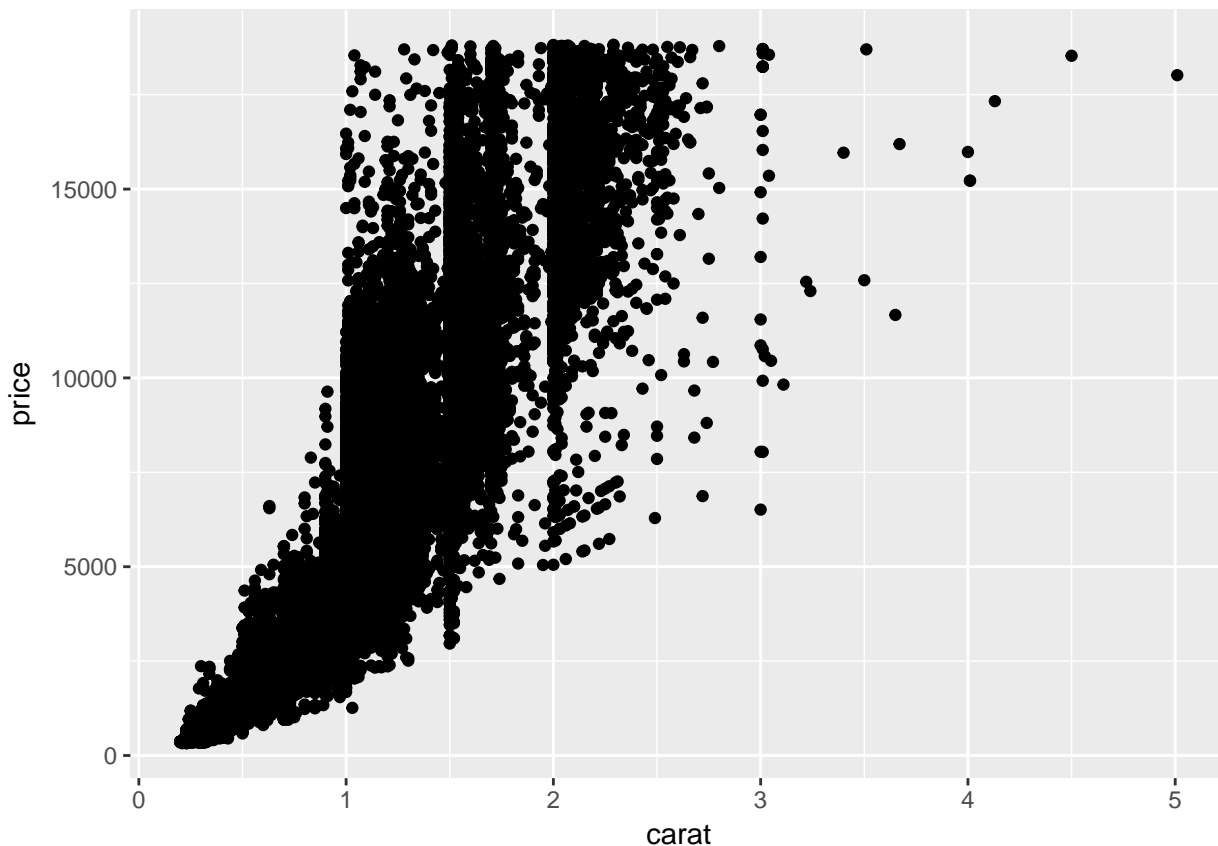
5.4.1 Instructions

- Explore the diamonds data frame with the `str()` function.
- Use the `+` operator to add `geom_point()` to the first `ggplot()` command. This will tell `ggplot2` to draw points on the plot.
- Use the `+` operator to add `geom_point()` and `geom_smooth()`. These just stack on each other! `geom_smooth()` will draw a smoothed line over the points.

```
# Explore the diamonds data frame with str()
str(diamonds)

## Classes 'tbl_df', 'tbl' and 'data.frame':  53940 obs. of  10 variables:
## $ carat : num  0.23 0.21 0.23 0.29 0.31 0.24 0.24 0.26 0.22 0.23 ...
## $ cut   : Ord.factor w/ 5 levels "Fair"<"Good"<...: 5 4 2 4 2 3 3 3 1 3 ...
## $ color : Ord.factor w/ 7 levels "D"<"E"<"F"<"G"<...: 2 2 2 6 7 7 6 5 2 5 ...
## $ clarity: Ord.factor w/ 8 levels "I1"<"SI2"<"SI1"<...: 2 3 5 4 2 6 7 3 4 5 ...
## $ depth : num  61.5 59.8 56.9 62.4 63.3 62.8 62.3 61.9 65.1 59.4 ...
## $ table : num  55 61 65 58 58 57 57 55 61 61 ...
## $ price : int  326 326 327 334 335 336 336 337 337 338 ...
## $ x     : num  3.95 3.89 4.05 4.2 4.34 3.94 3.95 4.07 3.87 4 ...
## $ y     : num  3.98 3.84 4.07 4.23 4.35 3.96 3.98 4.11 3.78 4.05 ...
## $ z     : num  2.43 2.31 2.31 2.63 2.75 2.48 2.47 2.53 2.49 2.39 ...
```

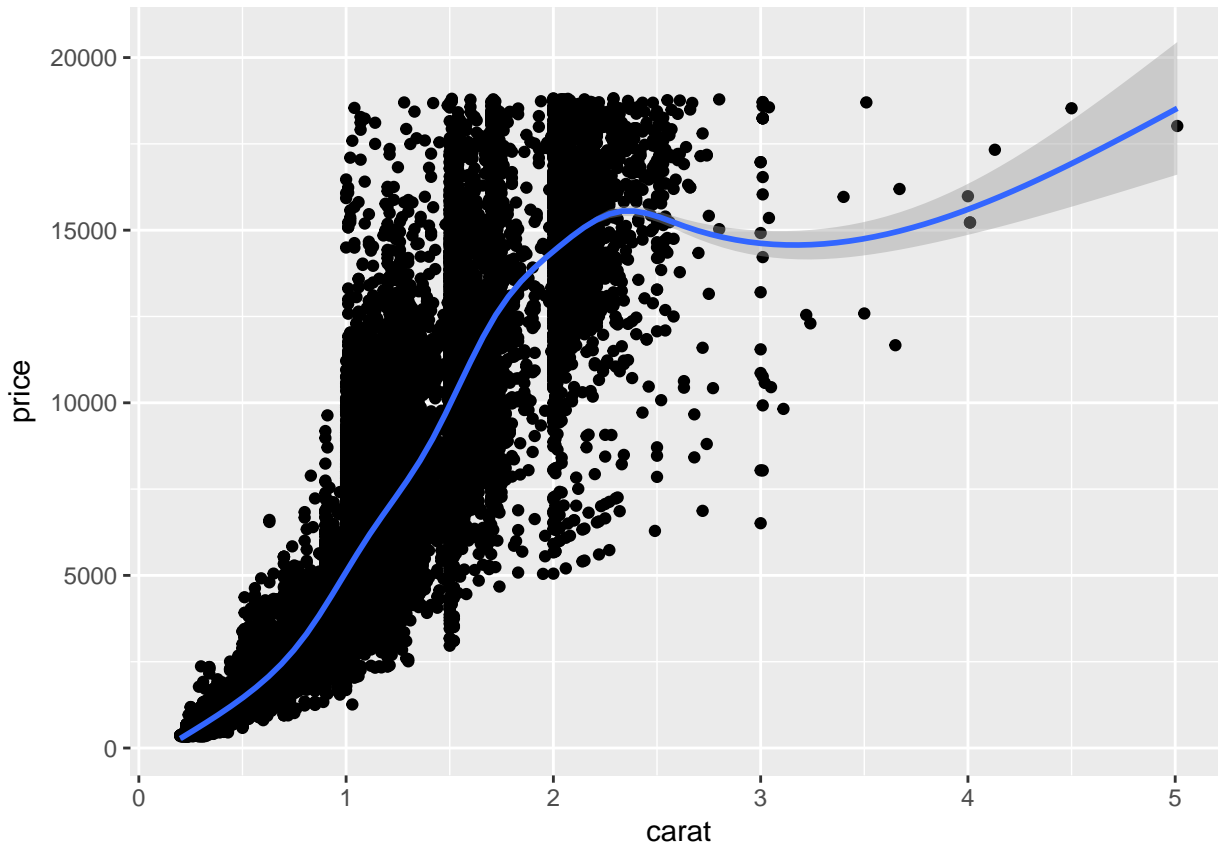
```
# Add geom_point() with +
ggplot(diamonds, aes(x = carat, y = price)) +
  geom_point()
```



```
# Add geom_point() and geom_smooth() with +
```

```
ggplot(diamonds, aes(x = carat, y = price)) +  
  geom_point() +  
  geom_smooth()
```

```
## `geom_smooth()` using method = 'gam'
```



```
# Lovely layering! If you had executed the command without adding a +, it would  
# produce an error message 'No layers in plot' because you are missing the third  
# essential layer - the geom layer.
```

5.5 Exploring ggplot2, part 5

The code for last plot of the previous exercise is available in the script on the right. It builds a scatter plot of the diamonds dataset, with carat on the x-axis and price on the y-axis. `geom_smooth()` is used to add a smooth line.

With this plot as a starting point, let's explore some more possibilities of combining geoms.

5.5.1 Instructions

- Plot 2 - Copy and paste plot 1, but show only the smooth line, no points.
- Plot 3 - Show only the smooth line, but color according to clarity by placing the argument `color = clarity` in the `aes()` function of your `ggplot()` call.
- Plot 4 - Draw translucent colored points.
 - Copy the `ggplot()` command from plot 3 (with clarity mapped to color).
 - Remove the smooth layer.

- Add the points layer back in.
- Set $\alpha = 0.4$ inside `geom_point()`. This will make the points 40% transparent.