

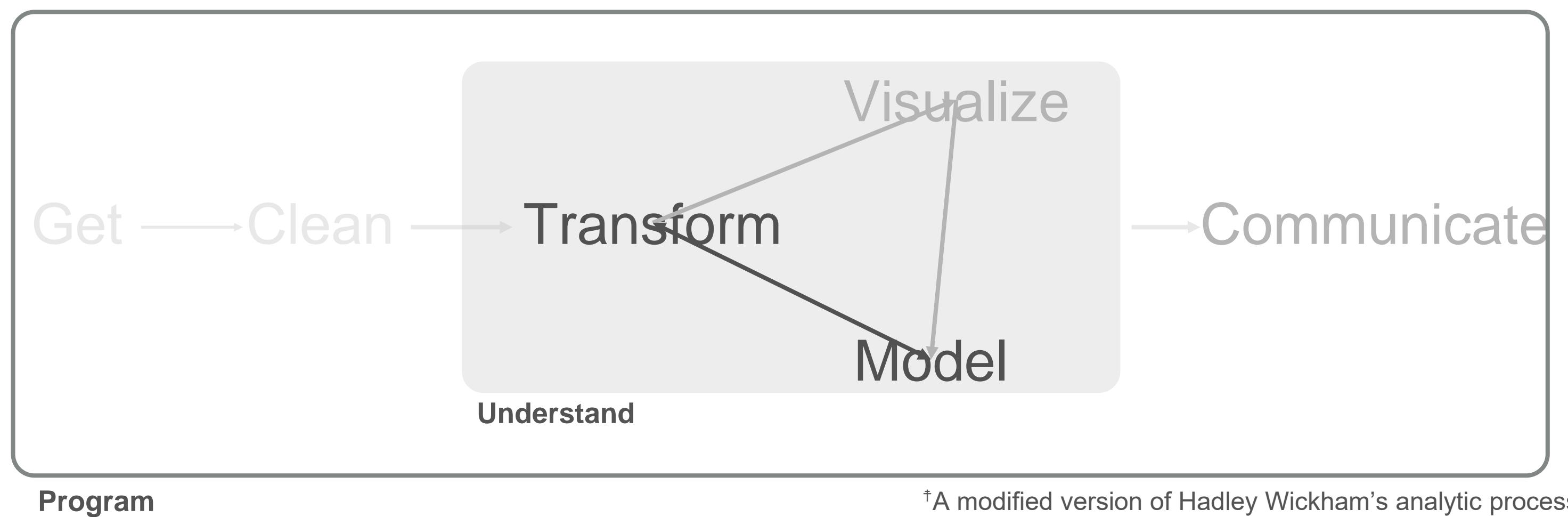
# TODAY'S CLASS

6:00PM – 7:30PM: if-else statements, for loops, **purrr** package

7:45PM – 9:15PM: Developing functions

9:15PM – 9:50PM: Time to work on your final project

# CONTROL STATEMENTS & ITERATION



“Great design is iteration of good design.”

– M. Cobanli

# CONTROL STATEMENTS & ITERATION

- Reducing code duplication has three main benefits:
  - It's easier to see the intent of your code, because your eyes are drawn to what's different, not what stays the same.
  - It's easier to respond to changes in requirements. As your needs change, you only need to make changes in one place, rather than remembering to change every place that you copied-and-pasted the code.
  - You're likely to have fewer bugs because each line of code is used in more places.

*Control statements and iteration functions provide two means to reduce code duplication*

# PREREQUISITES



# PACKAGE PREREQUISITE

```
library(tidyverse)
Loading tidyverse: ggplot2
Loading tidyverse: tibble
Loading tidyverse: tidyr
Loading tidyverse: readr
Loading tidyverse: purrr
Loading tidyverse: dplyr
Conflicts with tidy packages -----
filter(): dplyr, stats
lag():   dplyr, stats
```

# CONTROL STATEMENTS



# SYNTAX OF AN if STATEMENT

```
if (test_expression) {  
    statement  
}
```

Evaluates if the **test expression** is TRUE

If TRUE: execute **statement**

If FALSE: do nothing

# SYNTAX OF AN if STATEMENT

```
x <- c(8, 3, -2, 5)
```

```
if(x < 0) {  
  print("x contains a negative number")  
}
```

What do you expect to happen?

# SYNTAX OF AN if STATEMENT

```
x <- c(8, 3, -2, 5)  
  
if(x < 0) {  
  print("x contains a negative number")  
}  
  
Warning message:  
In if (x < 0) print("Negative") :  
  the condition has length > 1 and only the first  
  element will be used
```

Trick question - we actually get a warning

The if statement looks for a single conditional value. Here,  $x < 0$  will return a vector of 4 conditional values.

# SYNTAX OF AN if STATEMENT

```
x <- c(8, 3, -2, 5)
if(any(x < 0)) {
  print("x contains a negative number")
}
```

Instead, we can assess if **any** x values are less than 0

Now what do you expect to happen?

# SYNTAX OF AN if STATEMENT

```
x <- c(8, 3, -2, 5)
if(any(x < 0)) {
  print("x contains a negative number")
}
[1] "x contains a negative number"
```

Instead, we can assess if **any** x values are less than 0

Now what do you expect to happen?

# SYNTAX OF AN if STATEMENT

```
x <- c(8, 3, -2, 5)
if(any(x < 0)) {
  print("x contains a negative number")
}
[1] "x contains a negative number"
```

Instead, we can assess if **any** x values are less than 0

Now what do you expect to happen?

*Change -2 to 2. Now what happens when you run this code?*

# SYNTAX OF AN if...else STATEMENT

```
if (test_expression) {  
    statement 1  
} else {  
    statement 2  
}
```

We can extend an if statement with an else statement.

# SYNTAX OF AN if...else STATEMENT

```
x <- c(8, 3, -2, 5)
if(any(x < 0)) {
  print("x contains negative number(s)")
} else{
  print("x contains all positive numbers")
}
[1] "x contains negative number(s)"
```

Now, if the **test expression** is TRUE we print **statement 1** and if it is FALSE we print **statement 2**.

# SYNTAX OF AN if...else STATEMENT

```
x <- 7  
  
if(x >= 10) {  
  print("x exceeds acceptable tolerance levels")  
} else if(x >= 0 & x < 10) {  
  print("x is within acceptable tolerance  
levels")  
} else {  
  print("x is negative")  
}  
[1] "x is within acceptable tolerance levels"
```

And we can expand this further to include **multiple test expressions**.

# SYNTAX OF AN if...else STATEMENT

```
x <- 7  
  
if(x >= 10) {  
  print("x exceeds acceptable tolerance levels")  
} else if(x >= 0 & x < 10) {  
  print("x is within acceptable tolerance  
levels")  
} else {  
  print("x is negative")  
}  
[1] "x is within acceptable tolerance levels"
```

And we can expand this further to include multiple test expressions.

Pro-tip: else statements must go on the same line as the } or R will not read the else part of the if-else chain!

# YOUR TURN - PART 1

In your data folder you have many data sets. You want to create an if else control statement that will look to see if a particular “Month-XX.csv” file exists in this folder. For Part 1, build your if-else statement so that you create the appropriate file name based on a month input:

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 13  
  
if (month %in% 1:9) {  
  ...  
} else if (month %in% 10:12) {  
  ...  
} else {  
  ...  
}
```

If **month** equals 1-9 the output should look like “Month-08.csv”

If **month** equals 10-12 the output should look like “Month-12.csv”

If **month** is not 1-12 the output should look like “Invalid month”

# SOLUTION

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 1  
  
if (month %in% 1:9) {  
  paste0("data/", file_1, 0, month, file_2)  
} else if (month %in% 10:12) {  
  paste0("data/", file_1, month, file_2)  
} else {  
  print("Invalid month")  
}  
[1] "data/Month-01.csv"
```

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 13  
  
if (month %in% 1:9) {  
  paste0("data/", file_1, 0, month, file_2)  
} else if (month %in% 10:12) {  
  paste0("data/", file_1, month, file_2)  
} else {  
  print("Invalid month")  
}  
[1] "Invalid month"
```

## YOUR TURN - PART 2

*The file.exists function will check to see if a file exists at the given path.*

*Incorporate file.exists in your if-else statement so now the output provides a TRUE, FALSE, or “Invalid month” response. Is the December data (“Month-12.csv”) present?*

# SOLUTION

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 12  
  
if(month %in% 1:9) {  
  file_name <- paste0("data/", file_1, 0, month, file_2)  
  file.exists(file_name)  
} else if(month %in% 10:12) {  
  file_name <- paste0("data/", file_1, month, file_2)  
  file.exists(file_name)  
} else {  
  print("Invalid month")  
}  
[1] FALSE
```

# ITERATION

for loops



# SYNTAX OF A forLOOP

```
for(i in 1:100) {  
  <do stuff here with i>  
}
```

For each **element** in a **sequence**  
performs defined **tasks**

# SYNTAX OF A forLOOP

```
for(i in 1:100) {  
  <do stuff here with i>  
}
```

For each element in a sequence  
performs defined tasks

```
for(i in seq_along(x)) {  
  <do stuff here with i>  
}
```

Alternative - why use `seq_along?`

# SYNTAX OF A forLOOP

```
years <- 2010:2017
```

```
for (i in seq_along(years)) {  
  output <- paste("The year is", years[i])  
  print(output)  
}
```

What does this for loop do?

# SYNTAX OF A forLOOP

```
years <- 2010:2017
```

```
for (i in seq_along(years)) {  
  output <- paste("The year is", years[i])  
  print(output)  
}
```

```
[1] "The year is 2010"  
[1] "The year is 2011"  
[1] "The year is 2012"  
[1] "The year is 2013"  
[1] "The year is 2014"  
[1] "The year is 2015"  
[1] "The year is 2016"  
[1] "The year is 2017"
```

What does this for loop do?

# SYNTAX OF A forLOOP

```
result <- vector(mode = "character",
                  length = length(years))

for (i in seq_along(years)) {
  output <- paste("The year is", years[i])
  result[i] <- output
}

result
[1] "The year is 2010" "The year is 2011"
[3] "The year is 2012" "The year is 2013"
[5] "The year is 2014" "The year is 2015"
[7] "The year is 2016" "The year is 2017"
```

When saving results from a for loop we want to:

1. initiate the output outside of the for loop
2. save the results from the for loop to the output

# SYNTAX OF A forLOOP

```
x <- 7  
  
if(x >= 10) {  
  print("x exceeds acceptable tolerance levels")  
} else if(x >= 0 & x < 10) {  
  print("x is within acceptable tolerance  
levels")  
} else {  
  print("x is negative")  
}  
[1] "x is within acceptable tolerance levels"
```

Remember this extended if..else statement?

Let's implement it into a for loop

# SYNTAX OF A forLOOP

```
x <- c(-1, 7, 8, 11)
tolerance <- vector(mode = "character",
                      length = length(x))

for (i in seq_along(x)) {
  if(x[i] >= 10) {
    value <- "x exceeds acceptable tolerance levels"
  } else if(x[i] >= 0 & x[i] < 10) {
    value <- "x is within acceptable tolerance levels"
  } else {
    value <- "x is negative"
  }
  tolerance[i] <- value
}
```

1. Vector to analyze
2. Initiate output shell
3. For each element in our x
4. Assess test expressions, execute relevant statement, and save as value
5. For that respective element add the value in the corresponding element in out output shell

# YOUR TURN

*Remember this if-else statement we made?*

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 12  
  
if(month %in% 1:9) {  
  file_name <- paste0("data/", file_1, 0, month, file_2)  
  file.exists(file_name)  
} else if(month %in% 10:12) {  
  file_name <- paste0("data/", file_1, month, file_2)  
  file.exists(file_name)  
} else {  
  print("Invalid month")  
}
```

# YOUR TURN

*Develop a for loop that will loop through the months provided and import those .csv files that are present.*

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 1:13  
  
for(i in month) {  
  < magic happens here />  
}  
}
```

# YOUR TURN

*Develop a for loop that will loop through the months provided and import those .csv files that are present.*

```
file_1 <- "Month-"  
file_2 <- ".csv"  
month <- 1:13  
  
for(i in month) {  
  < magic happens here />  
}  
}
```

Run your for loop for months 1-13:

- If a particular month is available import it as “df.month.1”, “df.month.2”, ...
- If a particular month is not available, provide the response: “There is no data for month x”
- If a particular month is invalid (i.e. 13), provide the response: “x is an invalid month”

```

file_1 <- "Month-"
file_2 <- ".csv"
month <- 1:13

for(i in month) {

  # create file name
  if(i %in% 1:9) {
    file_name <- paste0("data/", file_1, 0, i, file_2)
  } else if(i %in% 10:12) {
    file_name <- paste0("data/", file_1, i, file_2)
  } else {
    response <- paste(i, "is an invalid month")
    print(response)
    next
  }

  # import data
  if(file.exists(file_name)) {
    df <- read_csv(file_name)
    assign(paste0("df.month.", i), df)
    rm(df)
  } else {
    response <- paste("There is no available data for month", i)
    print(response)
  }
}

```

# SOLUTION

1. Create file names.
2. If invalid month, provide response and skip to next iteration.
3. If the file exists, import and rename.
4. If the file does not exist for a given month, provide a response

```

file_1 <- "Month-"
file_2 <- ".csv"
month <- 1:13

for(i in month) {

  # create file name
  if(i %in% 1:9) {
    file_name <- paste0("data/", file_1, 0, i, file_2)
  } else if(i %in% 10:12) {
    file_name <- paste0("data/", file_1, i, file_2)
  } else {
    response <- paste(i, "is an invalid month")
    print(response)
    next
  }

  # import data
  if(file.exists(file_name)) {
    df <- read_csv(file_name)
    assign(paste0("df.month.", i), df)
    rm(df)
  } else {
    response <- paste("There is no available data for month", i)
    print(response)
  }
}

```

# SOLUTION

When you run this for months 1:13

The screenshot shows the RStudio interface with the 'Environment' tab selected. The 'Data' section lists 13 data frames: df.month.1, df.month.10, df.month.11, df.month.2, df.month.3, df.month.4, df.month.5, df.month.6, df.month.7, df.month.8, df.month.9, df.month.12, and df.month.13. Each entry shows the number of observations (54535 to 94315) and variables (10) for that specific month.

df.month.i	54535 obs. of 10 variables
df.month.1	54535 obs. of 10 variables
df.month.10	80277 obs. of 10 variables
df.month.11	94315 obs. of 10 variables
df.month.2	44380 obs. of 10 variables
df.month.3	53259 obs. of 10 variables
df.month.4	51033 obs. of 10 variables
df.month.5	55079 obs. of 10 variables
df.month.6	59666 obs. of 10 variables
df.month.7	64268 obs. of 10 variables
df.month.8	69492 obs. of 10 variables
df.month.9	71855 obs. of 10 variables
df.month.12	
df.month.13	

and for months 12 & 13 the response is:

```
[1] "There is no available data for month 12"
[1] "13 is an invalid month"
```

```

file_1 <- "Month-"
file_2 <- ".csv"
month <- 1:13

for(i in month) {

  # create file name
  if(i %in% 1:9) {
    file_name <- paste0("data/", file_1, 0, i, file_2)
  } else if(i %in% 10:12) {
    file_name <- paste0("data/", file_1, i, file_2)
  } else {
    response <- paste(i, "is an invalid month")
    print(response)
  }
}

# read in all the data frames
df <- lapply(month, function(i) {
  read.csv(paste0("data/", i), df)
})

# loop through each month and write to file
for(i in month) {
  if(i %in% 1:9) {
    file_name <- paste0("data/", file_1, 0, i, file_2)
  } else if(i %in% 10:12) {
    file_name <- paste0("data/", file_1, i, file_2)
  } else {
    response <- paste("There is no available data for month", i)
    print(response)
  }
}

```

# SOLUTION

When you run this for months 1-12

	Env	Variables
1	df.month.1	51033 obs. of 10 variables
2	df.month.2	55079 obs. of 10 variables
3	df.month.3	59666 obs. of 10 variables
4	df.month.4	64268 obs. of 10 variables
5	df.month.5	69492 obs. of 10 variables
6	df.month.6	71855 obs. of 10 variables
7	df.month.7	
8	df.month.8	
9	df.month.9	
10	df.month.10	
11	df.month.11	
12	df.month.12	
13	df.month.13	

Since all our data frames have similar variables (we just get updated data each month), how could we change this so you just create one single data frame?

and for months 12 & 13 the response is:

```
[1] "There is no available data for month 12"
[1] "13 is an invalid month"
```

```

file_1 <- "Month-"
file_2 <- ".csv"
month <- 1:13

# create empty data frame
df.all.months <- data.frame(NULL)

for(i in month) {

  # create file name
  if(i %in% 1:9) {
    file_name <- paste0("data/", file_1, 0, i, file_2)
  } else if(i %in% 10:12) {
    file_name <- paste0("data/", file_1, i, file_2)
  } else {
    response <- paste(i, "is an invalid month")
    print(response)
    next
  }

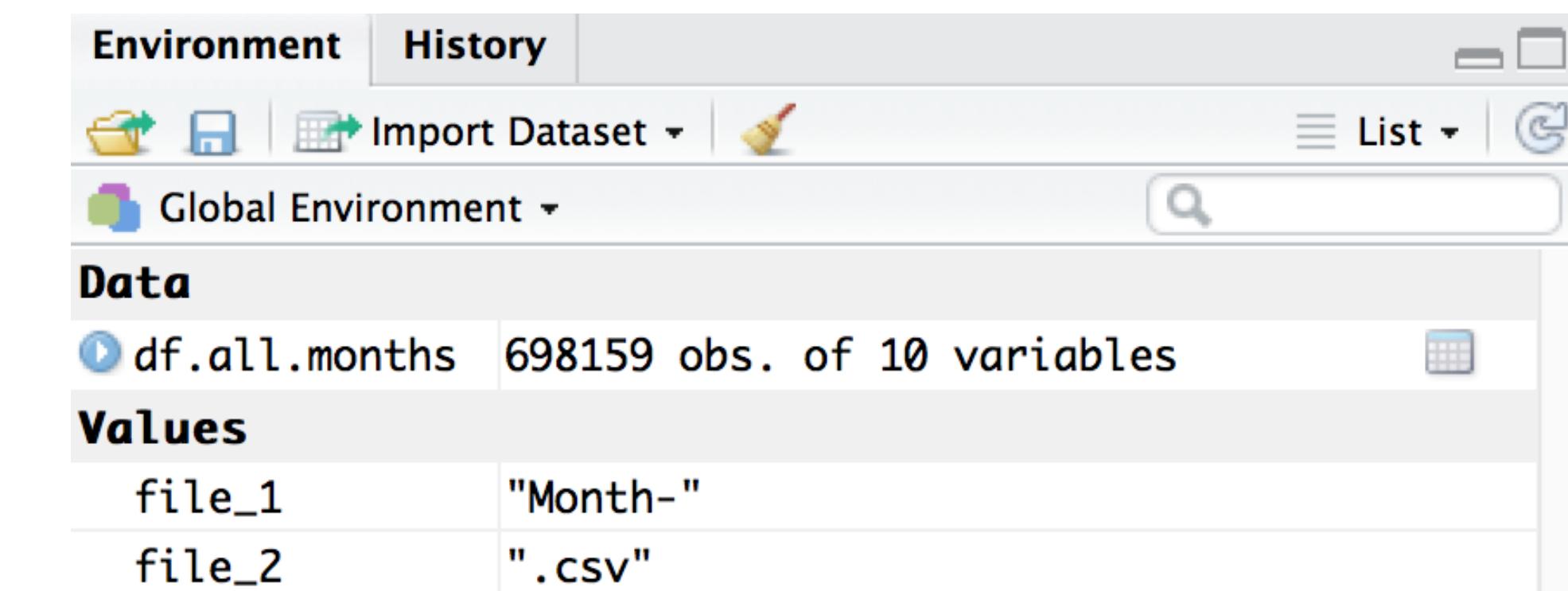
  # import data
  if(file.exists(file_name)) {
    df <- read.csv(file_name)
    df.all.months <- rbind(df.all.months, df)
    rm(df)
  } else {
    response <- paste("There is no available data for month", i)
    print(response)
  }
}

```

# SOLUTION

1. Create an empty data frame.

2. Then as you import the files, you can rbind the new data to our empty data frame.



# ITERATION

map functions



# INTRODUCING purrr

The **purrr** package provides an alternative means to iterate over elements and perform a set function

Each map function works the same `map_*(x, f)`:

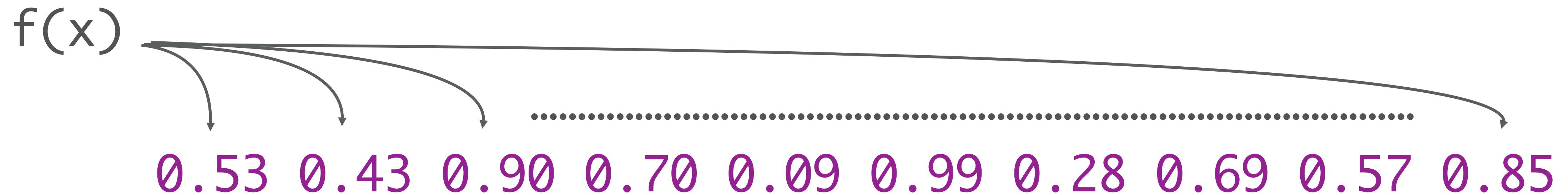
1. loop over `x`
2. apply function `f` to each element
3. return the results

This is an alternative approach to the `apply` family of functions



# BASICS OF THE MAP FUNCTIONS

`map_*(vector, function)`



# BASICS OF THE MAP FUNCTIONS

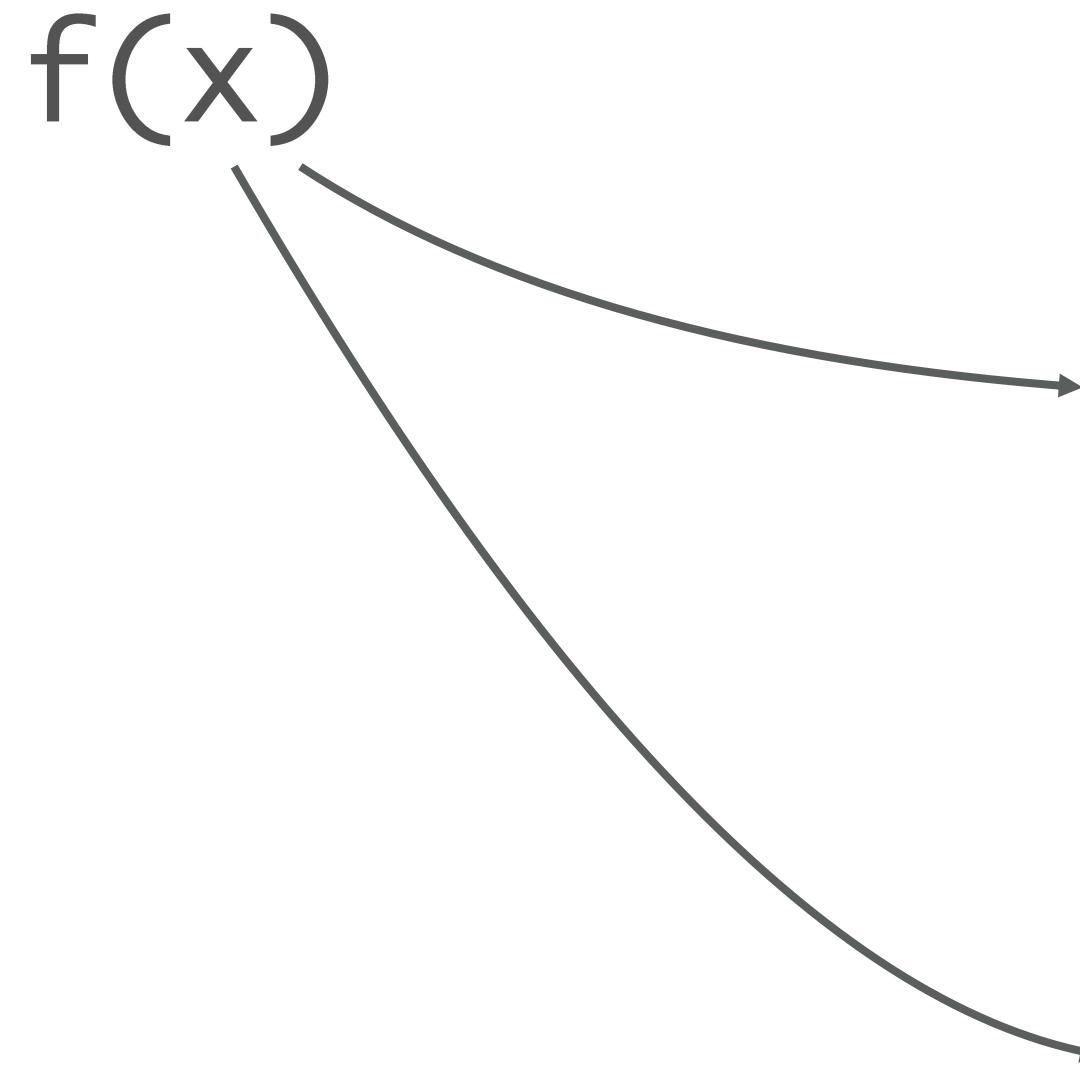
```
map_*(data.frame, function)
```

$f(x)$

		mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Mazda	RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
Mazda	RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
Datsun	710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
Hornet	4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
Hornet	Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
Valiant		18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
Duster	360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
Merc	240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2

# BASICS OF THE MAP FUNCTIONS

```
map_*(list, function)
```



```
[,1] [,2] [,3] [,4] [,5]  
[1,] 0.1843049 0.05852987 0.2002625 0.2860744 0.7361251  
[2,] 0.3005814 0.77366238 0.3798870 0.5418013 0.1937544
```

```
[,1] [,2] [,3] [,4] [,5]  
[1,] 0.9181339 0.6694026 0.3084275 0.4940026 0.08608652  
[2,] 0.5877923 0.5218446 0.9769865 0.6842613 0.04770349
```

# BASICS OF THE MAP FUNCTIONS

The `map_*` function we use is determined by the output of the function

- `map()` returns a list
- `map_dbl()` returns a double vector
- `map_lgl()` returns a logical vector
- `map_int()` returns a integer vector
- `map_chr()` returns a character vector

# BASICS OF THE MAP FUNCTIONS

```
map_dbl(mtcars, mean)
mtcars %>% map_dbl(mean)

  mpg          cyl        disp        hp
20.090625  6.187500 230.721875 146.687500
  drat          wt        qsec        vs
  3.596563  3.217250 17.848750  0.437500
  am          gear        carb
  0.406250  3.687500  2.812500
```

```
mtcars %>% map(mean)
$mpg
[1] 20.09062
$cyl
[1] 6.1875
```

The `map_*` functions incorporate the `%>%` operator

# YOUR TURN!

*With the iris data set, use the map functions to answer these three questions:*

- 1. What is the class of each variable?*
- 2. What is the mean value for each variable?*
- 3. Which variables have a mean value greater than 5?*

# SOLUTION

```
# what is the class of each variable?
```

```
iris %>% map_chr(class)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
"numeric"	"numeric"	"numeric"	"numeric"	"factor"	

```
# what is the mean value for each variable?
```

```
iris %>% map_dbl(mean)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
	5.843333	3.057333	3.758000	1.199333	NA

# SOLUTION

```
# which variables have a mean value greater than 5?
```

```
# option 1
```

```
iris %>%  
  map_dbl(mean) %>%  
  map_lgl(~ . > 5)
```

```
# option 2
```

```
iris %>% map_lgl(~ mean(.) > 5)
```

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
TRUE	FALSE	FALSE	FALSE	NA

# ADVANTAGES OF THE MAP FUNCTIONS

- Handy shortcuts for specifying `.f`
- More consistent than `sapply()`, `lapply()`, which makes them better for programming
- Takes much less time to solve iteration problems (because they are written in C)

# SHORTCUTS FOR SPECIFYING `f`

```
map(df, summary)
```

An existing function

```
map(df, myfunction)
```

An existing function you created

```
map(df, function(x) sum(is.na(x)))
```

An anonymous function defined on the fly

```
map(df, ~ sum(is.na(.)))
```

An anonymous function defined using a formula shortcut

# SHORTCUTS FOR SPECIFYING. f

*Can you find what variables in the nycflights13::flights data have missing values and how many missing values they have?*

# SHORTCUTS FOR SPECIFYING `f`

*Can you find what variables in the `nycflights::flights` data have missing values and how many missing values they have?*

	year	month	day	dep_time	sched_dep_time
dep_delay	0	0	0	8255	0
flight	8255	arr_time	sched_arr_time	arr_delay	carrier
distance	0	tailnum	origin	9430	0
hour	2512	0	0	dest	air_time
minute	0	0	0	0	9430
time_hour	0	0	0	0	0

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames...*

```
cyl <- split(mtcars, mtcars$cyl)
str(cyl)
List of 3
 $ 4:'data.frame': 11 obs. of  11 variables:
 ..$ mpg : num [1:11] 22.8 24.4 22.8 32.4 30.4 33.9 21.5 27.3 26 30.4 ...
 ..$ cyl : num [1:11] 4 4 4 4 4 4 4 4 4 4 ...
 ..$ disp: num [1:11] 108 146.7 140.8 78.7 75.7 ...
 ..$ hp   : num [1:11] 93 62 95 66 52 65 97 66 91 113 ...
 ..$ drat: num [1:11] 3.85 3.69 3.92 4.08 4.93 4.22 3.7 4.08 4.43 3.77 ...
 ..$ wt   : num [1:11] 2.32 3.19 3.15 2.2 1.61 ...
 ..$ qsec: num [1:11] 18.6 20 22.9 19.5 18.5 ...
 ..$ vs   : num [1:11] 1 1 1 1 1 1 1 1 0 1 ...
 ..$ am   : num [1:11] 1 0 0 1 1 1 0 1 1 1 ...
 ..$ gear: num [1:11] 4 4 4 4 4 4 3 4 5 5 ...
```

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames*

cyl[[1]]	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames*

```
mtcars %>%  
  split(. $cyl) %>%  
  map(~ lm(mpg ~ wt, data = .))  
$ `4`
```

```
Call:  
lm(formula = mpg ~ wt, data = .)
```

Coefficients:

(Intercept)	wt
39.571	-5.647

```
$ `6`  
Call:  
lm(formula = mpg ~ wt, data = .)
```

1. Apply regression model over each data frame

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames*

```
mtcars %>%  
  split(. $cyl) %>%  
  map(~ lm(mpg ~ wt, data = .)) %>%  
  map(summary)  
$`4`
```

Call:

```
lm(formula = mpg ~ wt, data = .)
```

Residuals:

Min	1Q	Median	3Q	Max
-4.1513	-1.9795	-0.6272	1.9299	5.2523

1. Apply regression model over each data frame
2. Access results with the summary function

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames*

```
mtcars %>%  
  split(.\$cyl) %>%  
  map(~ lm(mpg ~ wt, data = .)) %>%  
  map(summary) %>%  
  map_dbl("r.squared")  
    4          6          8  
0.5086326 0.4645102 0.4229655
```

1. Apply regression model over each data frame
2. Access results with the summary function
3. Get specific results by subsetting each list item

# MAP FUNCTIONS FOR QUICK MODEL COMPARISONS

*Say we want to apply a model over a list of data frames*

```
mtcars %>%  
  split(.\$cyl) %>%  
  map(~ lm(mpg ~ wt, data = .)) %>%  
  map(summary) %>%  
  map("coefficients") %>%  
  map_dbl(2)  
    4           6           8  
-5.647025 -2.780106 -2.192438
```

1. Apply regression model over each data frame
2. Access results with the summary function
3. Get specific results by subsetting each list item

# YOUR TURN!

1. *split the ggplot2::diamonds data set by cut*
2. *run a regression on each list item `lm(price ~ carat, data = .)`*
3. *get the summary of the regression*
4. *how does the "r.squared" compare across models?*
5. *how do the model slopes (aka “coefficients”) compare?*

# SOLUTION

## *Compare R<sup>2</sup> values*

```
ggplot2::diamonds %>%
  split(.\$cut) %>%
  map(~ lm(price ~ carat, data = .)) %>%
  map(summary) %>%
  map_dbl("r.squared")
  Fair      Good Very Good Premium     Ideal
0.7383940 0.8509539 0.8581622 0.8556336 0.8670887
```

# SOLUTION

## *Compare slopes values*

```
ggplot2::diamonds %>%
  split(.\$cut) %>%
  map(~ lm(price ~ carat, data = .)) %>%
  map(summary) %>%
  map("coefficients") %>%
  map_dbl(2)
```

	Fair	Good	Very Good	Premium	Ideal
5924.495	7479.636	7935.972	7807.752	8192.391	

SO LITTLE TIME!



*If you catch yourself copy and pasting the same code procedure then there is a good chance that you can incorporating some kind of iteration procedure...and I will be looking for this!*



# WHAT TO REMEMBER

# FUNCTIONS TO REMEMBER

Operator/Function	Description
<code>if, if...else, ifelse</code>	Conditional control statements
<code>for, while, repeat</code>	Looping control statements
<code>seq_along</code>	Sequencing argument
<code>break, next</code>	Arguments to exit or skip a loop iteration
<code>map, map dbl, map int, map chr, map lgl</code>	Family of iteration functions
<code>split</code>	Split a data frame based on categorical variable levels