

# Review of Basic R Objects and Tidyverse

# Let's Get Started

## Objects!!

- What are R objects?
- Why are they important?
- Can you name some object types?
  - vectors
  - data frames
  - lists
  - functions
  - operators
- How do you create an object (e.g., a vector)?
  - The assignment operator (`<-` or `=`) and the `c()` function 😊

```
x <- c(lots, of, cool, stuff)
```

# R Objects

## Vectors

What type of vector does each produce?

```
## Vectors
x = c(9,4,5,2,6)
y = c("male", "female", "female", "male", "female")
z = factor(y)
```

1. x: numeric
2. y: character
3. z: factor

# R Objects

## Data Frames and Lists

What is a `data.frame`? What about a `list`?

```
df = data.frame(  
  x = c(9,4,5,2,6),  
  y = c("male", "female",  
        "female", "male",  
        "female"),  
  z = factor(y)  
)
```

```
l1 = list(  
  x = c(9,4,5,2,6),  
  y = c("male", "female",  
        "female", "male",  
        "female"),  
  z = factor(y)  
)
```

画卷 A spreadsheet-like table and is the core data object that you will use.

包裹 It's much like a bag that you can put any object in (i.e., a flexible version of a `data.frame`).

Note: Data frames are a special list where all the elements are the same size.

# Importing Data

Almost any type of data can be imported into R with little effort.

We will show:

- CSV
- tab delimited
- space delimited
- SPSS
- SAS export

Others are possible 😊

# Importing Data

## Base R

### CSV

```
df <- read.csv("File.csv")
```

### Tab Delimited

```
df <- read.table("File.txt",  
                  header = TRUE,  
                  sep = "\t")
```

### Space Delimited

```
df <- read.table("File.txt",  
                  header = TRUE,  
                  sep = " ")
```

## Others

### CSV

```
library(readr)  
df <- read_csv("File.csv")
```

### SPSS

```
library(haven)  
df <- read_spss("File.sav")
```

### SAS

```
library(foreign)  
df <- read.xport("File.xpt")
```

# R Objects

## Functions 💪

It's how we do stuff in R!

What are the pieces of a function (a named function)?

```
myfunction = function(arguments){  
  stuff = that(the, function, does)  
  stuff  
}
```

1. a name
2. **function()**
3. arguments
4. stuff inside the {}
5. a return value (usually)

# Functions 💪

- How do you create your own function?

**Write your own function that gives you the mean and range of each variable in a data frame.**

**Now update that function so it only takes the mean and range of numeric variables.**

- One way:

```
mean_range = function(df){  
  means = lapply(df, mean)  
  mins = lapply(df, min)  
  maxs = lapply(df, max)  
  final = data.frame("Mean" = unlist(means),  
                     "Range" = unlist(maxs) - unlist(mins))  
  final  
}
```

# Functions 💪

```
mean_range = function(df){  
  means = lapply(df, mean)  
  mins = lapply(df, min)  
  maxs = lapply(df, max)  
  final = data.frame("Mean" = unlist(means),  
                     "Range" = unlist(maxs) - unlist(mins))  
  final  
}
```

- What does each piece do?
  - `lapply()`
  - `unlist()`
  - `final`
- What does `mean_range()` return?

# Functions 💪

```
mean_range = function(df){  
  means = lapply(df, mean)  
  mins = lapply(df, min)  
  maxs = lapply(df, max)  
  final = data.frame("Mean" = unlist(means),  
                     "Range" = unlist(maxs) - unlist(mins))  
  final  
}
```

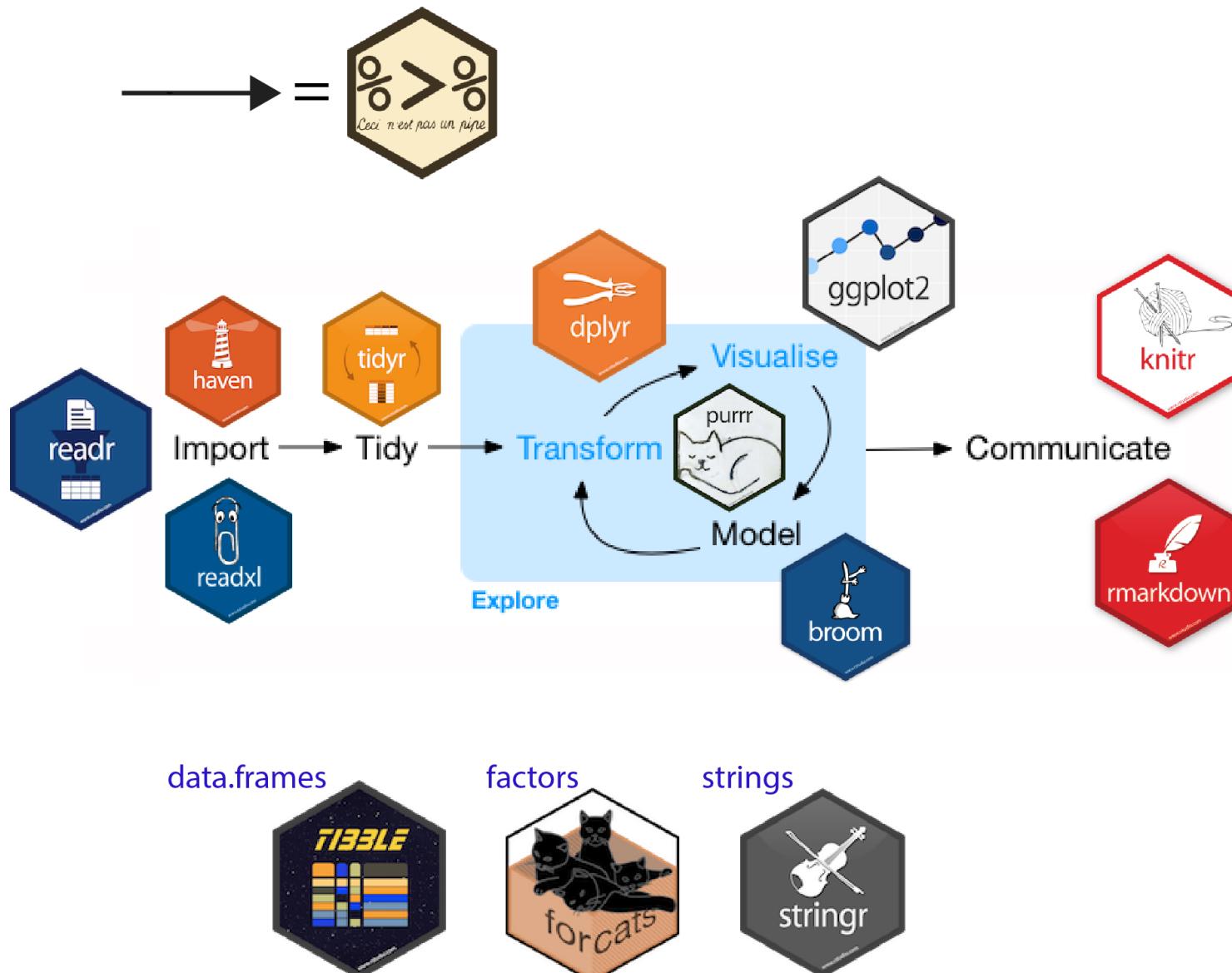
- What does each piece do?
  - `lapply()`
  - `unlist()`
  - `final`
- What does `mean_range()` return?

**What is something you'd like R to do for you? (let's create a function for it)**

# Questions about R Objects?

# Questions about R Objects?

## Now onto Tidyverse



# Tidyverse

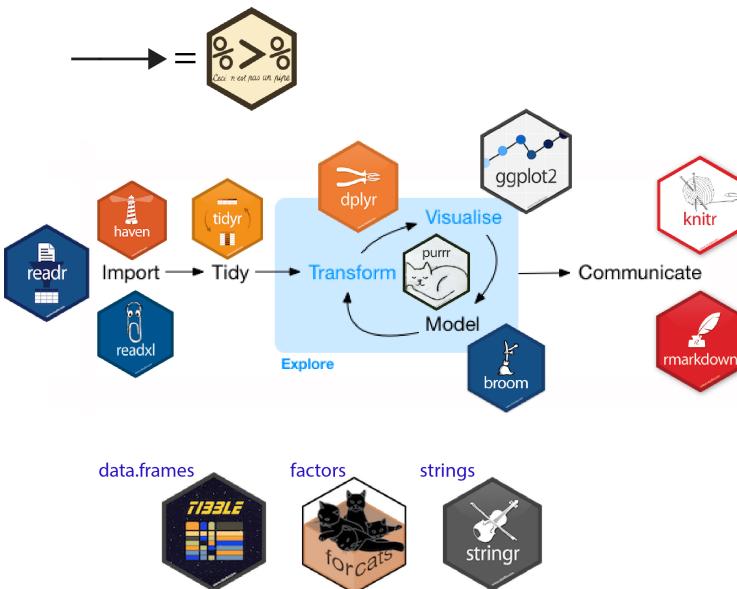
A group of packages that make working with data intuitive and clean.

## The steps of tidyverse:

1. import
2. tidy
3. transform
4. visualize
5. model
6. communicate

All these have the same overall goal:

- **to produce "tidy" data.**



# Tidy Data

Tidy data:

1. column is a variable
2. row is observation

The tidy form depends on many features of the data set.

E.g.

- For longitudinal data, this is generally the long form.

ID	X	Y	Z
1	.	.	.
1	.	.	.
2	.	.	.
2	.	.	.
...	.	.	.

# Tidy Data

We often need to reshape the data to get it into this form:

```
library(furniture)
df_long = long(df,
               c("var1", "var2")
               v.names = "var",
               idvar = "id")
```

`long()` takes a wide form to long form (which in many cases is the "tidy" form)

ID	X	Y	Z
1	.	.	.
1	.	.	.
2	.	.	.
2	.	.	.
...	.	.	.

# Selecting, Filtering, Grouping, Summarizing

## Base R

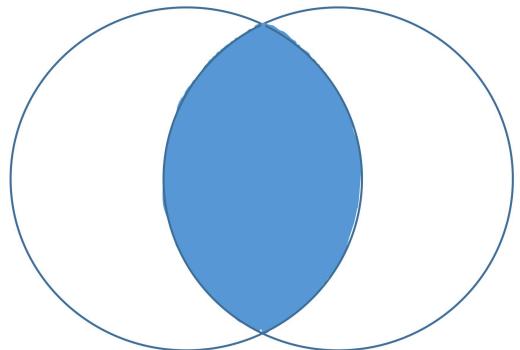
```
## Selecting  
d[, c("var1", "var2")]  
  
## Filtering  
d[d$var1 == 1, ]  
  
## Summarizing  
lapply(d, mean)  
  
## Summarizing by Groups  
tapply(d$var1, d$group, mean)
```

## Tidyverse

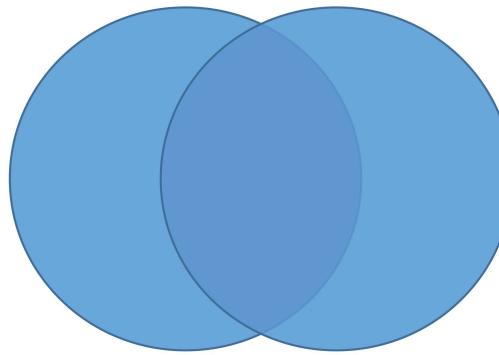
```
## Selecting  
select(d, var1, var2)  
  
## Filtering  
filter(d, var1 == 1)  
  
## Summarizing  
map(d, mean)  
  
## Summarizing by Groups  
d %>%  
  group_by(group) %>%  
  summarize(mean(var1))
```

# Join

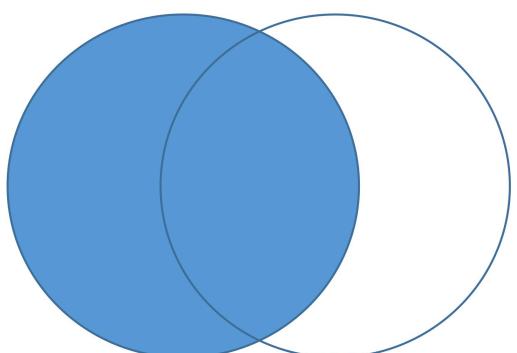
Inner Join



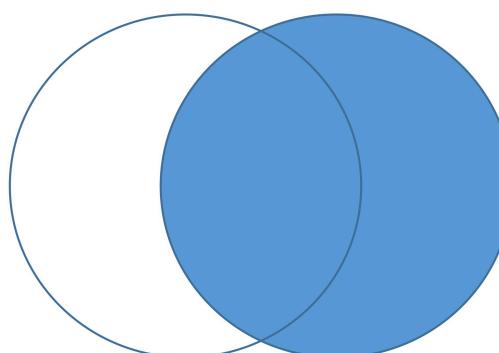
Full Join



Left Join



Right Join



# Join

From the tidyverse, you can easily do each type of join.

```
joined = d1 %>%
  inner_join(d2, by = "ID")
```

```
joined = d1 %>%
  left_join(d2, by = "ID")
```

```
joined = d1 %>%
  full_join(d2, by = "ID")
```

```
joined = d1 %>%
  right_join(d2, by = "ID")
```

Can do several in-a-row:

```
joined = d1 %>%
  full_join(d2, by = "ID") %>%
  full_join(d3, by = "ID") %>%
  full_join(d4, by = "ID")
```

Long and Wide

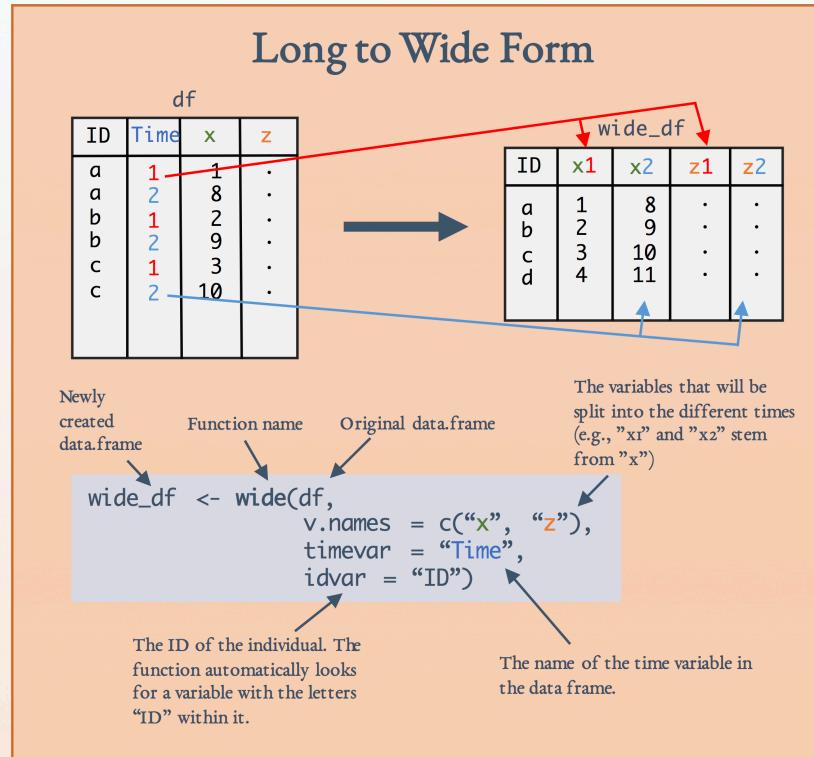
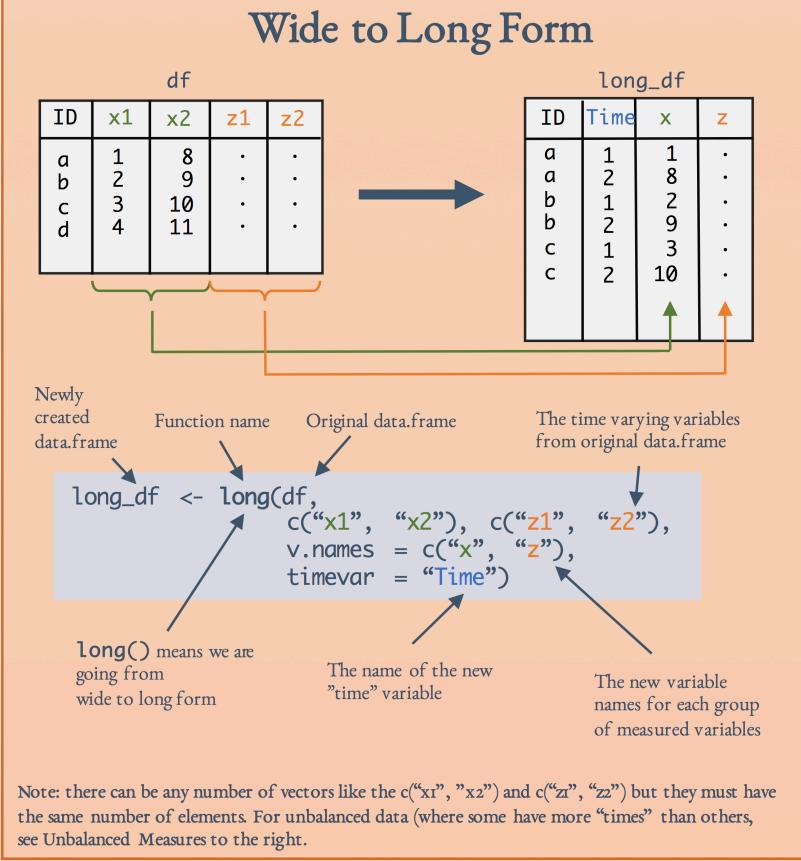
# Data Reshaping in R

Using the `furniture` package's `long()` and `wide()` functions

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Changing data from wide to long format is an essential tool in most data analyses. Long ("tidy") form is generally needed for plotting, statistical and other analyses.

Wide Format					Long Format				
ID	x1	x2	z1	z2	ID	Time	x	z	
a	1	8	.	.	a	1	1	.	
b	2	9	.	.	a	2	8	.	
c	3	10	.	.	b	1	2	.	
d	4	11	.	.	b	2	9	.	
					c	1	3	.	
					c	2	10	.	



### Unbalanced Measures

If the various measures have differing number of measurement points ("times"), then use the placeholder "miss". Say, our "x" measure had only one time point but our "z" measure had two, this code would work.

```
long(df,
      c("x1", "miss"),
      c("z1", "z2"),
      v.names = c("x", "z"),
      timevar = "time")
```

#### Additional Notes Regarding Data Reshaping:

- Recommendation:** Save only one data set (in wide or long format) that is your "main" data set—to avoid confusion about what different data sets contain. Then, use these reshaping functions in R to manipulate the data to get it in the form needed to plot and analyze. Finally, save the code ("syntax") and any output (you can save the reshaped data but it is not necessary).
- Term Definitions:** *measure* implies any distinct variable in the data set (e.g., `x2`), *time* implies anything that distinguishes the observations (e.g., time, cluster, location), *time varying variable* is any variable that has or could have a different value at each observation occasion.

# Long

Using the wide data below:

```
df_wide = data.frame(  
  "ID" = 1:10,  
  "X1" = rnorm(10),  
  "X2" = rnorm(10),  
  "Y1" = runif(10),  
  "Y2" = runif(10)  
)
```

We are going to reshape this to long format.

```
library(furniture)  
df_long = long(df_wide,  
  c("X1", "X2"),  
  c("Y1", "Y2"),  
  v.names = c("X", "Y"))  
  
## id = ID
```

# Long

Search:

ID	time	X	Y
1	1	0.1217	0.5385
2	1	-0.439	0.1527
3	2	0.8042	0.313
4	2	-0.1661	0.4858
5	3	0.4463	0.2052

Showing 1 to 5 of 20 entries

Previous

1

2

3

4

Next

# Wide

Using the `df_long` we just created, let's reshape it back to wide.

```
df_wide = wide(df_long,
                 v.names = c("X", "Y"),
                 timevar = "time")

## id = ID
```

Search:

ID	X.1	Y.1	X.2	Y.2
1	0.1217	0.5385	-0.439	0.1527
2	0.8042	0.313	-0.1661	0.4858
3	0.4463	0.2052	-0.7307	0.8848
4	0.229	0.008	1.1266	0.852
5	1.8668	0.9512	-1.5487	0.7041

Showing 1 to 5 of 10 entries

Previous

1

2

Next 23 / 24

# End of Review of Data Manipulation

Hopefully much of this was familiar.

Next we will review tables and ggplot2