

Importing and working with data in RStudio

Advanced Psychological Research Methods

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By the end of this section, you will be able to:

- Import data into R from excel, SPSS and csv files
- Identify different data structures and variable types
- Convert variables from one type to another
- Order, filter and group data
- Summarise data
- Create new variables from data

The *Tidyverse* set of packages

- A ‘toolkit’ of packages that are very useful for organising and manipulating data
- We will use the *haven* package to import SPSS files
- We will use the *dplyr* to organise data
- Also includes the *ggplot2* and *tidyR* packages which we will use later

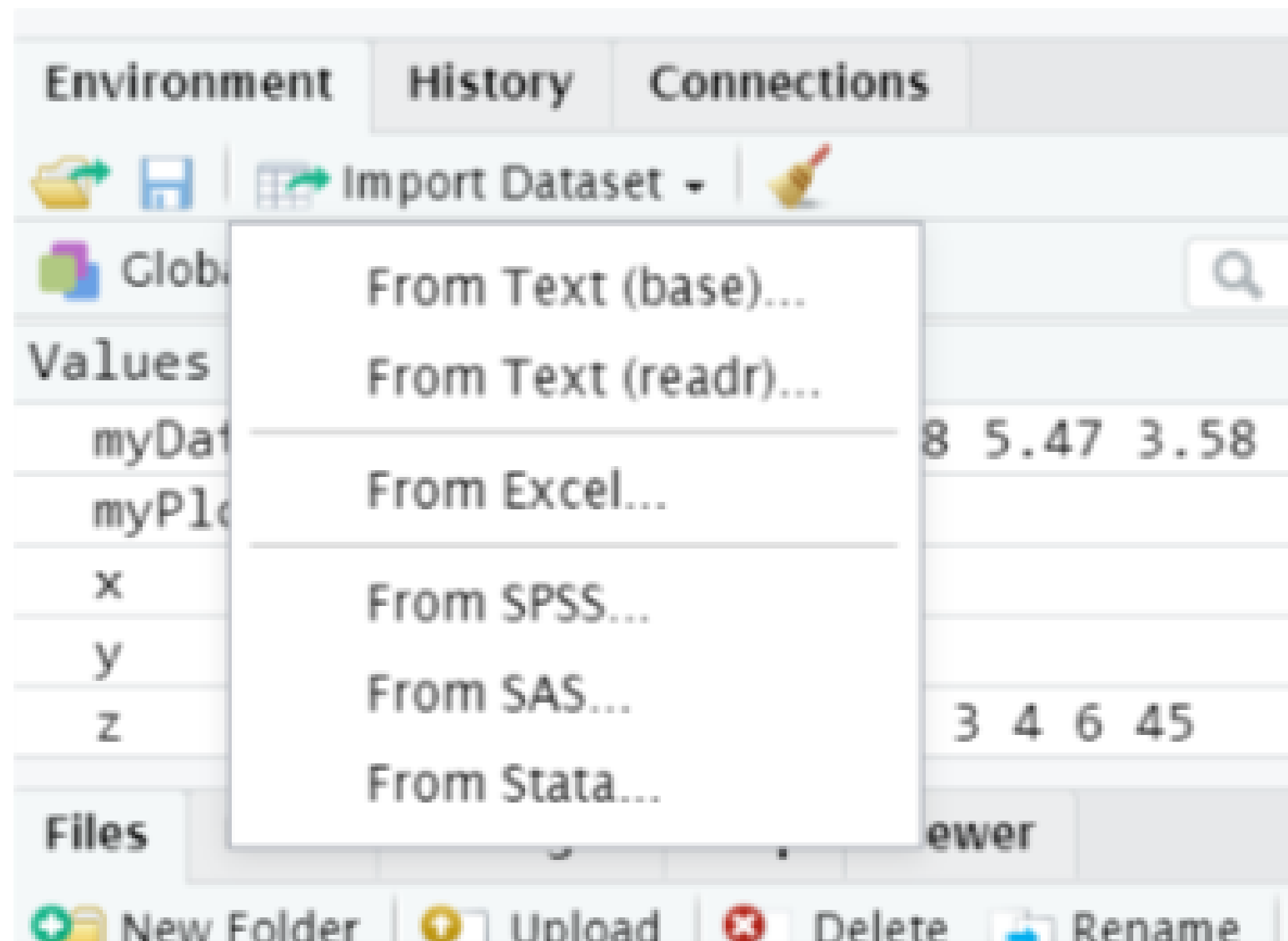
To install:

```
install.packages("tidyverse")
```

Importing .csv data into R

- Can use typed commands or the menu:

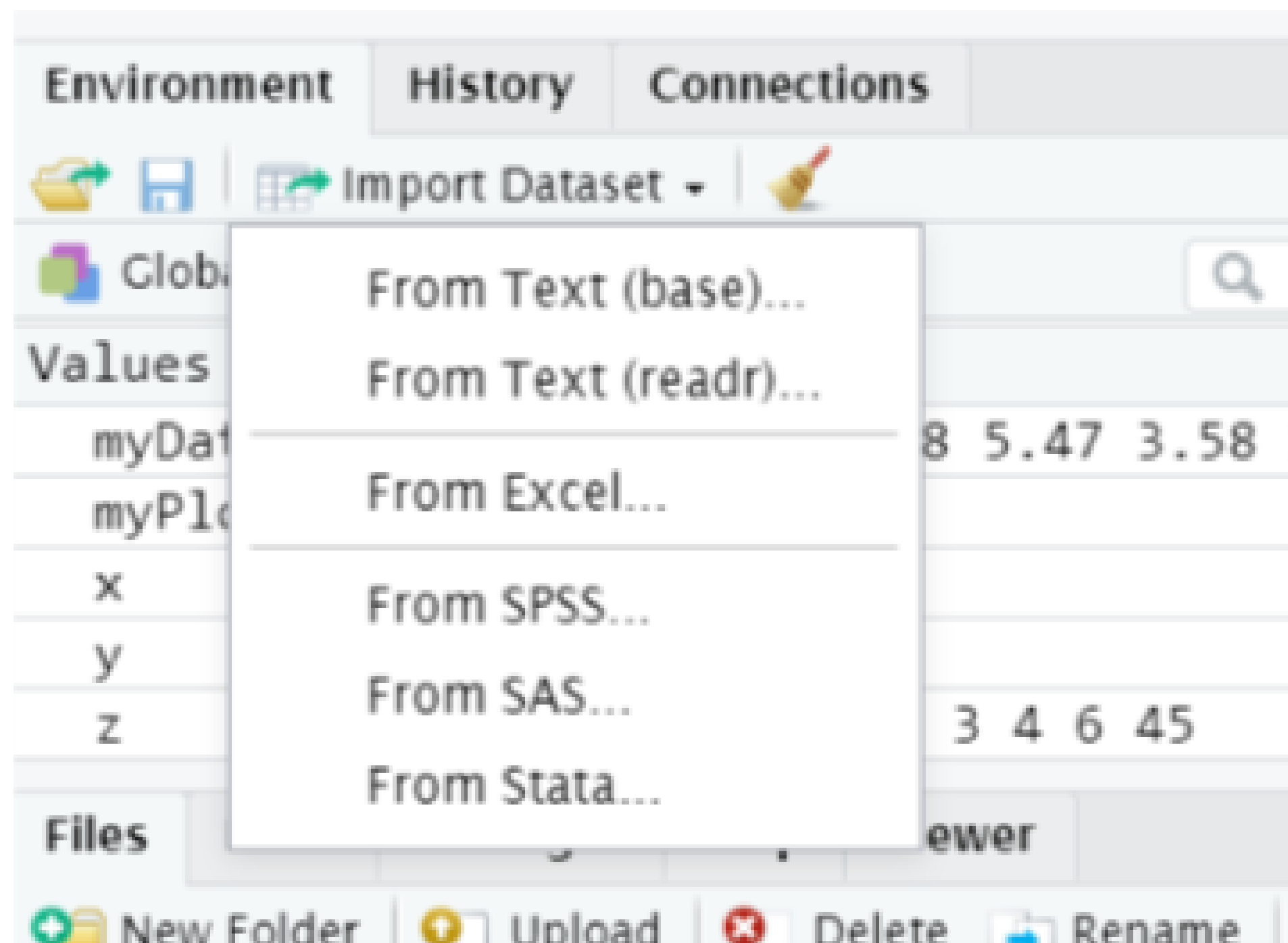
```
1 library(readr)
2 studentData <- read_csv("Datasets/studentData.csv")
```



Importing SPSS data files into R

- Can use typed commands or the menu:

```
1 library(haven)
2 mySPSSData <- read_sav("mySPSSFile.sav")
```



R can store many different data types

- Vectors: One-dimensional
- Data frames: Two-dimensional
- Matrices: Two-dimensional
- Arrays, Lists etc...

```
> v=c(3,4)
> A*v
      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
[1,]    3   33   63   93  123  153  183  213  243  273
[2,]    8   48   88  128  168  208  248  288  328  368
[3,]    9   39   69   99  129  159  189  219  249  279
[4,]   16   56   96  136  176  216  256  296  336  376
[5,]   15   45   75  105  135  165  195  225  255  285
[6,]   24   64  104  144  184  224  264  304  344  384
[7,]   21   51   81  111  141  171  201  231  261  291
[8,]   32   72  112  152  192  232  272  312  352  392
[9,]   27   57   87  117  147  177  207  237  267  297
[10,]  40   80  120  160  200  240  280  320  360  400
```

A data matrix:

The data frame

- A data frame is like a table or a two-dimensional array or matrix
- Each column contains values of one variable
- Each row contains one set of values
- Each column name must be unique

...1	route	grades	hoursOfStudy	hasDepdendants	satisfactionLevel
1	FullTime	56	3	Yes	Very
2	FullTime	47	1	Yes	Not at all
3	FullTime	72	8	Yes	Not at all
4	FullTime	79	0	Yes	Very
5	FullTime	79	4	Yes	Somewhat
6	FullTime	80	3	Yes	Somewhat

Checking the structure of the data

- The `str()` command will allow us to check how our data is structured:

```
1 str(studentData)
```

```
spec_tbl_df [100 × 6] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
 $ ..1 : num [1:100] 1 2 3 4 5 6 7 8 9 10 ...
 $ route : chr [1:100] "FullTime" "FullTime" "FullTime" "FullTime" ...
 $ grades : num [1:100] 56 47 72 79 79 80 76 39 85 41 ...
 $ hoursOfStudy : num [1:100] 3 1 8 0 4 3 1 6 2 5 ...
 $ hasDepdendants : chr [1:100] "Yes" "Yes" "Yes" "Yes" ...
 $ satisfactionLevel: chr [1:100] "Very" "Not at all" "Not at all" "Very" ...
- attr(*, "spec")=
 .. cols(
 ..   ..1 = col_double(),
 ..   route = col_character(),
 ..   grades = col_double(),
 ..   hoursOfStudy = col_double(),
 ..   hasDepdendants = col_character(),
 ..   satisfactionLevel = col_character()
 .. )
```

Notice that some of the variable types are incorrect

Changing variables from one data type to another

Collecting data – main levels of data

- There are four different levels of numerical data:

Nominal	Ordinal	Interval	Ratio
<ul style="list-style-type: none">• Categories• Can be counted• Cannot be ranked• Cannot be measured• Male/Female. Old/Young, Yes/No	<ul style="list-style-type: none">• Ranks• Can be counted• Can be ranked• Cannot be measured• 1st, 2nd, 3rd	<ul style="list-style-type: none">• Scale with exact values• Can be counted• Can be ranked• Can be measured• Can go below zero• E.g. temperature or difference score	<ul style="list-style-type: none">• Scale with exact values• Can be counted• Can be ranked• Can be measured• <u>Cannot</u> go below zero• E.g. A real number (time, count)

```
1 studentData$route <- as.factor(studentData$route)
2 studentData$hasDepdendants <- as.factor(studentData$hasDepdendants)
3 studentData$satisfactionLevel <- as.ordered(studentData$satisfactionLevel)
```

The code above changes the variables so that R recognises them as factors (nominal) or ordered factors (ordinal) variables

Changing variables from one data type to another

#2

Let's check the structure again:

```
1 str(studentData)
```

```
spec_tbl_df [100 × 6] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
 $ ..1 : num [1:100] 1 2 3 4 5 6 7 8 9 10 ...
 $ route : Factor w/ 2 levels "FullTime","PartTime": 1 1 1 1 1 1 1 1 1 1 ...
 $ grades : num [1:100] 56 47 72 79 79 80 76 39 85 41 ...
 $ hoursOfStudy : num [1:100] 3 1 8 0 4 3 1 6 2 5 ...
 $ hasDependants : Factor w/ 2 levels "No","Yes": 2 2 2 2 2 2 2 2 2 2 ...
 $ satisfactionLevel: Ord.factor w/ 3 levels "Not at all"<"Somewhat"<...: 3 1 1 3 2 2 3 2 1 2 ...
- attr(*, "spec")=
 .. cols(
 ..   ..1 = col_double(),
 ..   route = col_character(),
 ..   grades = col_double(),
 ..   hoursOfStudy = col_double(),
 ..   hasDependants = col_character(),
 ..   satisfactionLevel = col_character()
 .. )
```

```
1 str(studentData$satisfactionLevel)
```

```
Ord.factor w/ 3 levels "Not at all"<"Somewhat"<...: 3 1 1 3 2 2 3 2 1 2 ...
```

Changing variables from one data type to another

#3

Let's give a proper name to the row ID:

```
1 #Overwriting studentData with a new version, where the ID column has been correctly named
2 studentData <- studentData %>% rename(ID = ...1)
3 head(studentData) # viewing the first 5 rows of the data
```

```
# A tibble: 6 × 6
  ID route    grades hoursOfStudy hasDepdendants satisfactionLevel
<dbl> <fct>    <dbl>      <dbl>      <fct>      <ord>
1     1 FullTime    56          3 Yes        Very
2     2 FullTime    47          1 Yes        Not at all
3     3 FullTime    72          8 Yes        Not at all
4     4 FullTime    79          0 Yes        Very
5     5 FullTime    79          4 Yes        Somewhat
6     6 FullTime    80          3 Yes        Somewhat
```

Sorting data #1

- Using the *dplyr* package, we can arrange our data according to student grade:

```
1 arrange(studentData, grades)
```

```
# A tibble: 100 × 6
   ID route      grades hoursOfStudy hasDepdendants satisfactionLevel
  <dbl> <fct>    <dbl>      <dbl>    <fct>          <ord>
1    64 PartTime      35         7 No             Somewhat
2    93 PartTime      36         8 No             Somewhat
3    25 FullTime      37         7 Yes            Not at all
4    39 FullTime      37         2 Yes            Somewhat
5    46 FullTime      37         6 Yes            Very
6    61 PartTime      38         2 No             Somewhat
7     8 FullTime      39         6 Yes            Somewhat
8    31 FullTime      39         4 Yes            Somewhat
9    58 PartTime      39         7 No             Very
10   67 PartTime      40         6 No             Not at all
# ... with 90 more rows
# i Use `print(n = ...)` to see more rows
```

Sorting data #1

- Using the *dplyr* package, we can arrange our data according to student grade:

```
1 arrange(studentData, desc(grades)) # Arrange in descending order
```

```
# A tibble: 100 × 6
  ID route      grades hoursOfStudy hasDepdendants satisfactionLevel
  <dbl> <fct>    <dbl>      <dbl>    <fct>          <ord>
1     9 FullTime      85         2 Yes          Not at all
2    43 FullTime      85         0 Yes          Somewhat
3    59 PartTime      85         4 No           Very
4    76 PartTime      83         6 No          Not at all
5    48 FullTime      82         6 Yes          Not at all
6    68 PartTime      82         6 No           Very
7    71 PartTime      82         4 No          Somewhat
8    12 FullTime      81         8 Yes          Not at all
9    74 PartTime      81         5 No           Very
10     6 FullTime      80         3 Yes          Somewhat
# ... with 90 more rows
# i Use `print(n = ...)` to see more rows
```

Filtering data

- Show students who acheived a grade of less than 40%

```
1 filter(studentData, grades < 40)
```

A tibble: 9 × 6

	ID	route	grades	hoursOfStudy	hasDepdendants	satisfactionLevel
	<dbl>	<fct>	<dbl>	<dbl>	<fct>	<ord>
1	8	FullTime	39	6	Yes	Somewhat
2	25	FullTime	37	7	Yes	Not at all
3	31	FullTime	39	4	Yes	Somewhat
4	39	FullTime	37	2	Yes	Somewhat
5	46	FullTime	37	6	Yes	Very
6	58	PartTime	39	7	No	Very
7	61	PartTime	38	2	No	Somewhat
8	64	PartTime	35	7	No	Somewhat
9	93	PartTime	36	8	No	Somewhat



Filtering data #2

- Show part-time students who scored above 70%

```
1 filter(studentData, grades > 70 & route == "PartTime")
```

A tibble: 17 × 6

	ID	route	grades	hoursOfStudy	hasDepdendants	satisfactionLevel
	<dbl>	<fct>	<dbl>	<dbl>	<fct>	<ord>
1	52	PartTime	72	4	No	Very
2	59	PartTime	85	4	No	Very
3	60	PartTime	73	2	No	Very
4	62	PartTime	80	0	No	Somewhat
5	68	PartTime	82	6	No	Very
6	69	PartTime	74	7	No	Very
7	71	PartTime	82	4	No	Somewhat
8	73	PartTime	75	8	No	Very
9	74	PartTime	81	5	No	Very
10	75	PartTime	75	1	No	Very
11	76	PartTime	83	6	No	Not at all
12	78	PartTime	77	0	No	Very
13	81	PartTime	77	7	No	Somewhat



Using the “pipe” %>% to link steps of code

- part of the tidyverse package
- The pipe %>% allows us to:
 - write R commands in a way that is easier to read
 - Chain multiple commands together
- For example:

```
1 filteredData <- filter(studentData, grades > 70 & route == "PartTime")
2
3
4 filteredData <- studentData %>% filter(grades > 70 & route == "PartTime")
```

Grouping data

- It is possible to organise the data into groups and perform analysis on each group:

```
1 studentData %>% group_by(hasDepdendants) %>%  
2   summarise(mean = mean(grades), sd = sd(grades))
```

```
# A tibble: 2 × 3  
  hasDepdendants mean    sd  
  <fct>         <dbl> <dbl>  
1 No          62.4  14.3  
2 Yes         62.3  14.8
```

Remember: we can store that summary data as an object and call it later:

```
1 summaryTable <- studentData %>% group_by(hasDepdendants) %>%  
2   summarise(mean = mean(grades), sd = sd(grades))  
3  
4 summaryTable
```

```
# A tibble: 2 × 3  
  hasDepdendants mean    sd  
  <fct>      <dbl> <dbl>  
1 No         62.4   14.3  
2 Yes        62.3   14.8
```

Create new variables from data

We can create new variables from existing data using **mutate**

```
1 studentData %>% mutate(passFail = ifelse(grades > 40, "Pass", "Fail"))
```

```
# A tibble: 100 × 7
  ID route grades hoursOfStudy hasDepdendants satisfactionLevel passFail
<dbl> <fct> <dbl> <dbl> <fct> <ord> <chr>
1 1 FullTime 56 3 Yes Very Pass
2 2 FullTime 47 1 Yes Not at all Pass
3 3 FullTime 72 8 Yes Not at all Pass
4 4 FullTime 79 0 Yes Very Pass
5 5 FullTime 79 4 Yes Somewhat Pass
6 6 FullTime 80 3 Yes Somewhat Pass
7 7 FullTime 76 1 Yes Very Pass
8 8 FullTime 39 6 Yes Somewhat Fail
9 9 FullTime 85 2 Yes Not at all Pass
10 10 FullTime 41 5 Yes Somewhat Pass
# ... with 90 more rows
# i Use `print(n = ...)` to see more rows
```

Importing and Working with data in R – Tasks:

Importing and Working with data in R

0. Install the **tidyverse** package. When this has been done, enter the command **library("tidyverse")**
1. Use the **read.sav()** command to import the "**Datasets/salesData.sav**" file to a new object called **salesData**
2. Use the **str()** command to check the structure of **salesData** . There should be 5 variables: salary, married, numberOfVisits, age & valueOfSales
3. Change the variable **married** to a factor
4. Arrange the data from lowest to highest **sales** . What was the lowest **sales** value?
5. What is the mean **salary** ?
6. Filter the data to only display customers who are **married** (1 = married, 2 = not married)
7. Create a summary of the data to compare the mean and standard deviation of **sales** for married and non-married customers (1 = married, 2 = not married)
8. Create a new variable called **VIP** and label customers who spent over ?500 as "VIP" and other customers as "Non-VIP"