

# Data Import

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# Basic Data

- Data frames are the most flexible and one of the most used object type in R.
- There are many R functions to load, manipulate and save data frames.
- In RStudio you can use the import data wizard.
  - File | Import Dataset | From CSV ...
  - File | Import Dataset | From Excel ...
  - File | Import Dataset | From SPSS ...
  - File | Import Dataset | From SAS ...
  - File | Import Dataset | From Stata ...
- Alternatively can use R code.
- Using R code
  - Quicker to run.
  - Easy to share and for others to replicate

# Comma Separated Values

- Easiest way to import data is from files in the Comma Separated Values (CSV) format.
- A typical .csv file will look something like this.

```
"COUNTRY", "YEAR", "SAMPLE", "SERIAL", "GEOLEV1", "GEOLEV2", "PERNUM", "PERWT", "AGE", "NATI
591,1960,591196001,1000,591004,591004003,1,20,53,1,1,110
591,1960,591196001,1000,591004,591004003,2,20,54,1,1,120
591,1960,591196001,1000,591004,591004003,3,20,31,1,1,120
591,1960,591196001,1000,591004,591004003,4,20,22,1,2,212
591,1960,591196001,1000,591004,591004003,5,20,20,1,2,212
591,1960,591196001,1000,591004,591004003,6,20,16,1,2,212
591,1960,591196001,1000,591004,591004003,7,20,13,1,2,212
591,1960,591196001,1000,591004,591004003,8,20,5,1,0,0
591,1960,591196001,1000,591004,591004003,9,20,3,1,0,0
591,1960,591196001,1000,591004,591004003,10,20,2,1,0,0
591,1960,591196001,2000,591004,591004003,1,20,42,1,1,110
591,1960,591196001,3000,591004,591004003,1,20,58,1,1,110
591,1960,591196001,3000,591004,591004003,2,20,82,1,1,110
...
```

- CSV files can be viewed in Excel with commas removed
- Can convert a single Excel spreadsheet as a CSV file using the Save As option.

# Comma Separated Values

- The two most common ways to read CSV files into R are using:
  - `read.csv()` in the base package
  - `read_csv()` in the `readr` package.
- The `readr` package is part of the tidyverse set of packages (more on the tidyverse later)
- When data is read in they are different types of R objects
  - `data.frame` from `read.csv()`
  - `tbl_df` (tibble) from `read_csv()`
- They display differently when printed to the R console...

# Comma Separated Values

- Demonstrate loading data using R code with IPUMSI data
  - <http://international.ipums.org/international>
  - A large data base containing an census micro-data from around the world.
  - Can download CSV files (as well as Stata, SPSS and SAS)
  - Free. Registration required.
- The `file.show()` function opens files in their default program

```
> file.show("./data/ipumsi_pan1960.csv")
```

# Comma Separated Values

```
df0 <- read.csv(file = "../data/ipumsi_pan1960.csv")
df0
```

>	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT	AGE	NATIVITY
> 1	591	1960	591196001	1000	591004	591004003	1	20	53	1
> 2	591	1960	591196001	1000	591004	591004003	2	20	54	1
> 3	591	1960	591196001	1000	591004	591004003	3	20	31	1
> 4	591	1960	591196001	1000	591004	591004003	4	20	22	1
> 5	591	1960	591196001	1000	591004	591004003	5	20	20	1
> 6	591	1960	591196001	1000	591004	591004003	6	20	16	1
> 7	591	1960	591196001	1000	591004	591004003	7	20	13	1
> 8	591	1960	591196001	1000	591004	591004003	8	20	5	1
> 9	591	1960	591196001	1000	591004	591004003	9	20	3	1
> 10	591	1960	591196001	1000	591004	591004003	10	20	2	1
> 11	591	1960	591196001	2000	591004	591004003	1	20	42	1
> 12	591	1960	591196001	3000	591004	591004003	1	20	58	1
> 13	591	1960	591196001	3000	591004	591004003	2	20	82	1
> 14	591	1960	591196001	3000	591004	591004003	3	20	57	1
> 15	591	1960	591196001	4000	591004	591004003	1	20	62	1
> 16	591	1960	591196001	4000	591004	591004003	2	20	26	1
> 17	591	1960	591196001	5000	591004	591004003	1	20	42	1
> 18	591	1960	591196001	5000	591004	591004003	2	20	35	1
> 19	591	1960	591196001	5000	591004	591004003	3	20	17	1
> 20	591	1960	591196001	5000	591004	591004003	4	20	15	1
> 21	591	1960	591196001	5000	591004	591004003	5	20	8	1
> 22	591	1960	591196001	5000	591004	591004003	6	20	6	1

# Tibbles

```
> library(readr)
> df1 <- read_csv(file = "../data/ipumsi_pan1960.csv")
> df1
```

*# A tibble: 53,553 x 12*

	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT	AGE
	<int>	<int>	<int>	<int>	<int>	<int>	<int>	<int>	<int>
1	591	1960	591196001	1000	591004	591004003	1	20	53
2	591	1960	591196001	1000	591004	591004003	2	20	54
3	591	1960	591196001	1000	591004	591004003	3	20	31
4	591	1960	591196001	1000	591004	591004003	4	20	22
5	591	1960	591196001	1000	591004	591004003	5	20	20
6	591	1960	591196001	1000	591004	591004003	6	20	16
7	591	1960	591196001	1000	591004	591004003	7	20	13
8	591	1960	591196001	1000	591004	591004003	8	20	5
9	591	1960	591196001	1000	591004	591004003	9	20	3
10	591	1960	591196001	1000	591004	591004003	10	20	2

*# ... with 53,543 more rows, and 3 more variables: NATIVITY <int>,  
# EDATTAIN <int>, EDATTAIND <int>*

# Tibbles

- When you print `data.frames` you get everything.
  - If you are dealing with non-small data sets this is annoying
  - If you are dealing with very large data sets the printing can take a long time.
- A `tbl_df` is an improved `data.frame` with nice methods for high-level inspection.
  - By default `tbl_df` will print only the first 10 rows for large data sets
  - Will subdue extra columns that won't fit into your console
  - Provides the column type in a three letter abbreviations under the column names
  - Dimension information at the top.



# Comma Separated Values

- The `read_csv()` function can be much faster and avoids converting to factors (more on factors later).

## Useful arguments



```
a,b,c
1,2,3
4,5,NA
```

### Example file

```
write_csv(path = "file.csv",
x = read_csv("a,b,c\n1,2,3\n4,5,NA"))
```

A	B	C
1	2	3
4	5	NA

### No header

```
read_csv("file.csv",
col_names = FALSE)
```

x	y	z
A	B	C
1	2	3
4	5	NA

### Provide header

```
read_csv("file.csv",
col_names = c("x", "y", "z"))
```

1	2	3
4	5	NA

### Skip lines

```
read_csv("file.csv",
skip = 1)
```

A	B	C
1	2	3

### Read in a subset

```
read_csv("file.csv",
n_max = 1)
```

A	B	C
1	2	3
NA	NA	NA

### Missing Values

```
read_csv("file.csv",
na = c("4", "5", "."))
```

# Delimited Files

- The `read_csv()` function is a special case of `read_delim()`
- Different people and/or countries use different formats to separate values

a,b,c  
1,2,3  
4,5,NA

→

A	B	C
1	2	3
4	5	NA

**read\_csv()**  
Reads comma delimited files.  
*read\_csv("file.csv")*

a;b;c  
1;2;3  
4;5;NA

→

A	B	C
1	2	3
4	5	NA

**read\_csv2()**  
Reads Semi-colon delimited files.  
*read\_csv2("file2.csv")*

a|b|c  
1|2|3  
4|5|NA

→

A	B	C
1	2	3
4	5	NA

**read\_delim()**(*delim*, *quote* = "\"", *escape\_backslash* = FALSE, *escape\_double* = TRUE) Reads files with any delimiter.  
*read\_delim("file.txt", delim = "|")*

a b c  
1 2 3  
4 5 NA

→

A	B	C
1	2	3
4	5	NA

**read\_fwf()**(*col\_positions*)  
Reads fixed width files.  
*read\_fwf("file.fwf", col\_positions = c(1, 3, 5))*

**read\_tsv()**  
Reads tab delimited files. Also **read\_table()**.  
*read\_tsv("file.tsv")*

# Saving Data

- The two most common ways to write (save) CSV files into R are using:
  - `write.csv()` in the base package
  - `write_csv()` in the `readr` package.
- `write.csv()` will add a first column with the row name (usually a number)
- `write_csv()` saves just the data frame by default

```
> write_csv(x = df0, path = "./data/mynewfile.csv")
```

- `write_excel_csv()` works well with Chinese characters (can get lost with `write_csv()`)
- CSV files are the most common external data format to used with R.
- Users tend to save data as CSV, even if imported from a different format as they are small and simple.

# Exercise 1

- Open ex31.R and complete the following exercises. Once you have filled in, save the exercise file as "ex31.R".

```
# 0. Clear your workspace and set your working directory to your data folder in the
rm(list = ls())
setwd(dir = "C:/Users/Guy/Dropbox/APPI2017/exercise/data/")
##
##
##

# 1. Load the readr package

# 2. Open the "2010_Census_Populations_by_Zip_Code.csv" file using the file.show()

# 3. Use read_csv to read the data into R and call the results d1

# 4. Open the "unhcr_popstats_export_persons_of_concern.csv" file using the file.show()

# 5. a) Q: How many lines should we skip when reading the data?
#      A: 3
#      b) Q: Are there any missing values (if so how are they represented)?
#      A: *

# 6. Use read_csv to read the data into R and call the results d2
```

# Data Exploration

There are many R functions to explore your data.

Function	Description
<code>head()</code>	First rows of the data frame
<code>tail()</code>	Last rows of the data frame
<code>str()</code>	Structure of data frame
<code>summary()</code>	Summary of each column of the data frame
<code>dim()</code>	Dimensions of the data frame
<code>nrow()</code>	Number of rows in the data frame
<code>ncol()</code>	Number of columns in the data frame
<code>rownames()</code>	Row names in the data frame
<code>colnames()</code>	Column names in the data frame
<code>dimnames()</code>	Row and column names in the data frame
<code>View()</code>	Invoke a Data Viewer

# Data Exploration

```
> head(df0)
```

	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT	AGE
1	591	1960	591196001	1000	591004	591004003	1	20	53
2	591	1960	591196001	1000	591004	591004003	2	20	54
3	591	1960	591196001	1000	591004	591004003	3	20	31
4	591	1960	591196001	1000	591004	591004003	4	20	22
5	591	1960	591196001	1000	591004	591004003	5	20	20
6	591	1960	591196001	1000	591004	591004003	6	20	16

  

	NATIVITY	EDATTAIN	EDATTAIND
1	1	1	110
2	1	1	120
3	1	1	120
4	1	2	212
5	1	2	212
6	1	2	212

# Data Exploration

```
> str(df0)
'data.frame':   53553 obs. of  12 variables:
 $ COUNTRY   : int  591 591 591 591 591 591 591 591 591 591 ...
 $ YEAR      : int  1960 1960 1960 1960 1960 1960 1960 1960 1960 1960 ...
 $ SAMPLE    : int  591196001 591196001 591196001 591196001 591196001 591196001 591196001 591196001 591196001 591196001 ...
 $ SERIAL    : int  1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 ...
 $ GEOLEV1   : int  591004 591004 591004 591004 591004 591004 591004 591004 591004 591004 ...
 $ GEOLEV2   : int  591004003 591004003 591004003 591004003 591004003 591004003 591004003 591004003 591004003 591004003 ...
 $ PERNUM    : int  1 2 3 4 5 6 7 8 9 10 ...
 $ PERWT     : int  20 20 20 20 20 20 20 20 20 20 ...
 $ AGE       : int  53 54 31 22 20 16 13 5 3 2 ...
 $ NATIVITY  : int  1 1 1 1 1 1 1 1 1 1 ...
 $ EDATTAIN  : int  1 1 1 2 2 2 2 0 0 0 ...
 $ EDATTAIND : int  110 120 120 212 212 212 212 0 0 0 ...
```

# Data Exploration

```
> summary(df0)
```

COUNTRY		YEAR		SAMPLE		SERIAL	
Min.	:591	Min.	:1960	Min.	:591196001	Min.	: 1000
1st Qu.	:591	1st Qu.	:1960	1st Qu.	:591196001	1st Qu.	: 2883000
Median	:591	Median	:1960	Median	:591196001	Median	: 6135000
Mean	:591	Mean	:1960	Mean	:591196001	Mean	: 5974039
3rd Qu.	:591	3rd Qu.	:1960	3rd Qu.	:591196001	3rd Qu.	: 9063000
Max.	:591	Max.	:1960	Max.	:591196001	Max.	:11869000

  

GEOLEV1		GEOLEV2		PERNUM		PERWT	
Min.	:591002	Min.	:5.91e+08	Min.	: 1.000	Min.	:20
1st Qu.	:591004	1st Qu.	:5.91e+08	1st Qu.	: 2.000	1st Qu.	:20
Median	:591004	Median	:5.91e+08	Median	: 3.000	Median	:20
Mean	:591006	Mean	:5.91e+08	Mean	: 3.715	Mean	:20
3rd Qu.	:591008	3rd Qu.	:5.91e+08	3rd Qu.	: 5.000	3rd Qu.	:20
Max.	:591008	Max.	:5.91e+08	Max.	:28.000	Max.	:20

  

AGE		NATIVITY		EDATTAIN		EDATTAIND	
Min.	: 0.0	Min.	:1.00	Min.	:0.000	Min.	: 0.0
1st Qu.	: 7.0	1st Qu.	:1.00	1st Qu.	:0.000	1st Qu.	: 0.0
Median	:18.0	Median	:1.00	Median	:1.000	Median	:120.0
Mean	:22.7	Mean	:1.08	Mean	:1.016	Mean	:112.9
3rd Qu.	:35.0	3rd Qu.	:1.00	3rd Qu.	:1.000	3rd Qu.	:120.0
Max.	:98.0	Max.	:9.00	Max.	:9.000	Max.	:999.0



# Data Exploration

- In RStudio can also use the `View` function or click on the data frame in the Environment tab to (initially) view the first 1000 rows.
- Can filter and sort columns within the data view.

```
> View(df0)
```

# Factors

- When dealing with categorical data in R we often use factors
  - Special vectors that are character strings with an additional level attribute.
  - The level attribute provides further information on the order of the categorical data
- Older importing functions, such as `read.csv()`, will create factors by default when bringing data into R. (`stringsAsFactors = TRUE`)
  - Data importation functions in the tidyverse, such as `read_csv`, do not create factors by default
  - This is partly why they are faster at importing.
- Creating a factors can be done easily with the `factor()` function.
  - Levels follows alpha-numeric order by default.
- Rearranging the orders of the factors with the can be difficult. The `forcats` package has lots of helpful functions help work with factors and categorical data.
  - `fct_recode()` change the factor names
  - `fct_relevel()` reorder factors by hand
  - `fct_reorder()` reorder factors by sorting on a variable
  - `fct_inorder()` order factors by their appearance

# Factors

```
> # original numeric data
```

```
> table(df1$NATIVITY)
```

```
      1      2      9
49399  4132   22
```

```
> # covert to factor
```

```
> df1$NATIVITY <- factor(df1$NATIVITY)
```

```
> table(df1$NATIVITY)
```

```
      1      2      9
49399  4132   22
```

```
> # rename the levels with character strings
```

```
> library(forcats)
```

```
> df1$NATIVITY <- fct_recode(df1$NATIVITY, "native" = "1", "foreign" = "2", "missing"
```

```
> table(df1$NATIVITY)
```

```
native foreign missing
 49399   4132     22
```

```
> # reorder factors
```

```
> df1$NATIVITY <- fct_relevel(df1$NATIVITY, "foreign", "missing", "native")
```

```
> table(df1$NATIVITY)
```

```
foreign missing  native
  4132      22  49399
```

# Data Modes

- R can easily change the mode and object types of columns in data frames.

Function	Description
<code>as.numeric()</code>	creates a numeric vector
<code>as.character()</code>	creates a character vector
<code>as.integer()</code>	creates an integer vector
<code>as.factor()</code>	creates a factor vector
<code>as.data.frame()</code>	will turn vectors, matrices and lists into data frames

## Exercise 2 (ex32.R)

```
# 0. a) Check your working directory is in the course folder

#      b) Load the data and packages by sourcing the solution file for ex31.R
#####("./exercise-solutions/ex31.R")
##
##
##

# 1. What are the dimensions of d1 data frame object

# 2. Show a summary of each column of d1

# 3. Show the first three rows of d1
#      (Hint: Use head() function. See ?head to set the number of rows)

# 4. Show the column names of d1

# 5. Convert the `Year` column in d2 from integers to a character string

# 6. Convert the `Origin` column in d2 from a character string to a factor

# 7. Show the levels of Origin column

# 8. Convert the name column in d5 to a factor

# 9. Show the levels of name column
```

# Excel Data

- There is no functions in base R to read excel files.
- The readxl package has a read\_excel() function.
- In read\_excel() you can specify the sheet (either a name or number).
- To demonstrate we will use the SAPE18DT14.xls spreadsheet from the UK ONS on population estimates by age group in England and Wales output areas.

```
file.show("../data/SAPE18DT14.xls")
```

- Similar options to read\_csv() (na =, col\_names =, skip =)
- The guess\_max = argument can very useful for when dealing with long data frames
  - By default guess\_max = 1000, i.e. a guess of the data type will be based on the first 1000 rows.
  - Sometimes the guess might be wrong and throw a big red warning message - set guess\_max to a higher value - at the cost of speed

# Excel Data

```
> library(readxl)
> df2 <- read_excel(path = "./data/SAPE18DT14.xls", sheet = 2, skip = 3)
> df2
# A tibble: 7,549 x 23
  `Area Codes` `Area Names` X__1 `All ages` `0-4` `5-9`
    <chr>      <chr>      <chr>    <dbl> <dbl> <dbl>
1 E06000047 County Durham <NA> 519695 28446 28859
2 E02004297 <NA> County Durham 001 7912 455 421
3 E02004290 <NA> County Durham 002 5851 251 313
4 E02004298 <NA> County Durham 003 9858 488 524
5 E02004299 <NA> County Durham 004 8588 555 506
6 E02004291 <NA> County Durham 005 6957 427 376
7 E02004300 <NA> County Durham 006 7840 496 506
8 E02004292 <NA> County Durham 007 7845 350 444
9 E02004301 <NA> County Durham 008 9205 653 666
10 E02004302 <NA> County Durham 009 7766 508 469
# ... with 7,539 more rows, and 17 more variables: `10-14` <dbl>,
# `15-19` <dbl>, `20-24` <dbl>, `25-29` <dbl>, `30-34` <dbl>,
# `35-39` <dbl>, `40-44` <dbl>, `45-49` <dbl>, `50-54` <dbl>,
# `55-59` <dbl>, `60-64` <dbl>, `65-69` <dbl>, `70-74` <dbl>,
# `75-79` <dbl>, `80-84` <dbl>, `85-89` <dbl>, `90+` <dbl>
```

# Excel Data

- Saving R output as Excel file is not too easy... use CSV if you can.
- There are a number of older packages e.g. `xlsx` and `XLConnect` that are especially tricky.
  - Not too easy to install as require correct Java version.
  - Do have alternative options to read data from Excel.
- The `openxlsx` package does not require Java.
  - Must first define a workbook and sheet...



# Excel Data

```
> # install.packages("openxlsx")
> library(openxlsx)
> # create a empty workbook to fill
> wb0 <- createWorkbook(creator = "Guy")
> # create a empty sheet in the workbook
> addWorksheet(wb = wb0, sheetName = "small area population")
> # add your data
> writeData(wb = wb0, sheet = 1, x = df2)
> # add a filter and freeze the top row
> addFilter(wb = wb0, sheet = 1, rows = 1, cols = names(df2))
> freezePane(wb = wb0, sheet = 1, firstRow = TRUE)
>
> ## Save workbook to working directory
> saveWorkbook(wb0, file = "./data/xlexample.xlsx", overwrite = TRUE)
> file.show("./data/xlexample.xlsx")
```

## Exercise 3 (ex33.R)

```
# 0. Clear your workspace and set your working directory to your data folder in the  
  
##  
##  
##  
# 1. load the readxl package  
  
# 2. Use read_excel to read data on Male population totals in SAPE18DT14.xls into R  
  
# 3. Use read_excel to read data on ESTIMATES in WPP2015_FERT_F04_TOTAL_FERTILITY.xls  
  
# 4. Use read_excel to read data on MEDIUM VARIANT in WPP2015_POP_F07_1_POPULATION_B
```

# SPSS Data

- There are number of packages to import data from SPSS, Stata and SAS files.
- The two most popular are the `haven` and `foriegn` packages
- Each extends R `read()` and `write()` functions for different file types.
- The `haven` package is part of the tidyverse. It is bit newer, bit faster, outputs tibbles!

# SPSS Data

```
> library(haven)
> df3 <- read_sav(file = "./data/ipumsi_pan1960.sav")
> df3
```

*# A tibble: 53,553 x 12*

	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT
	<dbl+lbl>	<dbl+lbl>	<dbl+lbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	591	1960	591196001	1000	591004	591004003	1	20
2	591	1960	591196001	1000	591004	591004003	2	20
3	591	1960	591196001	1000	591004	591004003	3	20
4	591	1960	591196001	1000	591004	591004003	4	20
5	591	1960	591196001	1000	591004	591004003	5	20
6	591	1960	591196001	1000	591004	591004003	6	20
7	591	1960	591196001	1000	591004	591004003	7	20
8	591	1960	591196001	1000	591004	591004003	8	20
9	591	1960	591196001	1000	591004	591004003	9	20
10	591	1960	591196001	1000	591004	591004003	10	20

*# ... with 53,543 more rows, and 4 more variables: AGE <dbl+lbl>,  
# NATIVITY <dbl+lbl>, EDATTAIN <dbl+lbl>, EDATTAIND <dbl+lbl>*

# SPSS Data

- The labels for each data code do not display from `read_sav`
- However they are known to R.

```
> print_labels(df3$NATIVITY)
```

Labels:

value	label
0	NIU (not universe)
1	Native-born
2	Foreign-born
9	Unknown/missing

# SPSS Data

- Can change column to a characters using `as_factor()` function in the `haven` package

```
> # single column
> table(df3$NATIVITY)

 1      2      9
49399 4132   22

> df3$NATIVITY <- as_factor(df3$NATIVITY, "labels")
> table(df3$NATIVITY)

NIU (not universe)      Native-born      Foreign-born
                        0      49399      4132
Unknown/missing
22

> # all columns
> head(df3)
# A tibble: 6 x 12
  COUNTRY YEAR SAMPLE SERIAL GEOLEV1 GEOLEV2 PERNUM PERWT
<dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1     591  1960 591196001  1000  591004 591004003     1    20
2     591  1960 591196001  1000  591004 591004003     2    20
3     591  1960 591196001  1000  591004 591004003     3    20
4     591  1960 591196001  1000  591004 591004003     4    20
```

# SPSS Data

- The foreign package has limited support for newer SPSS formats (since 2000).
- For SPSS files will create a list by default (`to.data.frame = FALSE`)

```
> # install.packages("foreign")
> library(foreign)
> df4 <- read.spss(file = "./data/ipumsi_pan1960.sav", use.value.labels = FALSE, to
re-encoding from UTF-8
> head(df4, n = 2)
```

	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT	AGE
1	591	1960	591196001	1000	591004	591004003	1	20	53
2	591	1960	591196001	1000	591004	591004003	2	20	54

  

	NATIVITY	EDATTAIN	EDATTAIND
1	1	1	110
2	1	1	120

```
> df4 <- read.spss(file = "./data/ipumsi_pan1960.sav", to.data.frame = TRUE)
re-encoding from UTF-8
> head(df4, n = 2)
```

	COUNTRY	YEAR	SAMPLE	SERIAL	GEOLEV1	GEOLEV2	PERNUM	PERWT	AGE	
1	Panama	1960	Panama	1960	1000	591004	591004003	1	20	53
2	Panama	1960	Panama	1960	1000	591004	591004003	2	20	54

  

	NATIVITY	EDATTAIN	EDATTAIND	
1	Native-born	Less than primary	completed	No schooling
2	Native-born	Less than primary	completed	Some primary completed

# Stata Data

- The haven package works well for Stata files of all ages

```
> # haven
> df5 <- read_dta("../data/ipumsi_pan1960.dta")
> # gives a tibble
> head(df5, n = 3)
# A tibble: 3 x 12
```

	country	year	sample	serial	geolev1	geolev2	pernum	perwt
	<dbl+lbl>	<dbl+lbl>	<dbl+lbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	591	1960	591196001	1000	591004	591004003	1	20
2	591	1960	591196001	1000	591004	591004003	2	20
3	591	1960	591196001	1000	591004	591004003	3	20

```
# ... with 4 more variables: age <dbl+lbl>, nativity <dbl+lbl>,
# edattain <dbl+lbl>, edattaind <dbl+lbl>
> df5 <- as_factor(df5)
> head(df5, n = 3)
# A tibble: 3 x 12
```

	country	year	sample	serial	geolev1	geolev2	pernum	perwt	age
	<fctr>	<fctr>	<fctr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<fctr>
1	panama	1960	panama 1960	1000	591004	591004003	1	20	53
2	panama	1960	panama 1960	1000	591004	591004003	2	20	54
3	panama	1960	panama 1960	1000	591004	591004003	3	20	31

```
# ... with 3 more variables: nativity <fctr>, edattain <fctr>,
# edattaind <fctr>
```



# Stata Data

- For Stata > v12 `read.dta` in the `foreign` package will not work.

```
> # foreign... does not work
> df6 <- read.dta("./data/ipumsi_pan1960.dta")
> `Error in read.dta("./data/ipumsi_pan1960.dta") :
+   unable to open file: 'No such file or directory'`
```

# Stata Data

- The readstata13 package is a good alternative.
  - By default will use labels (generate.factors = TRUE)

```
> # install.packages("readstata13")
> library(readstata13)
> df6 <- read.dta13("./data/ipumsi_pan1960.dta", generate.factors = FALSE)
> head(df6, n = 3)
```

	country	year	sample	serial	geolev1	geolev2	pernum	perwt	age	
1	panama	1960	panama	1960	1000	591004	591004003	1	20	53
2	panama	1960	panama	1960	1000	591004	591004003	2	20	54
3	panama	1960	panama	1960	1000	591004	591004003	3	20	31

```

      nativity                                edattain                                edattaind
1 native-born less than primary completed                                no schooling
2 native-born less than primary completed some primary completed
3 native-born less than primary completed some primary completed
> df6 <- read.dta13("./data/ipumsi_pan1960.dta")
> head(df6, n = 3)
```

	country	year	sample	serial	geolev1	geolev2	pernum	perwt	age	
1	panama	1960	panama	1960	1000	591004	591004003	1	20	53
2	panama	1960	panama	1960	1000	591004	591004003	2	20	54
3	panama	1960	panama	1960	1000	591004	591004003	3	20	31

```

      nativity                                edattain                                edattaind
1 native-born less than primary completed                                no schooling
2 native-born less than primary completed some primary completed
```

# SAS data

- SAS files saved as .sas7bdat can work with haven read\_sas
- More difficult in foreign package read.ssd (not shown).

```
> # haven
> df7 <- read_sas("./data/ipumsi_pan1960.sas7bdat")
> # gives a tibble
> head(df7)
# A tibble: 6 x 12
  COUNTRY YEAR SAMPLE SERIAL GEOLEV1 GEOLEV2 PERNUM PERWT AGE
  <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1    591  1960 591196001   1000  591004 591004003     1    20    53
2    591  1960 591196001   1000  591004 591004003     2    20    54
3    591  1960 591196001   1000  591004 591004003     3    20    31
4    591  1960 591196001   1000  591004 591004003     4    20    22
5    591  1960 591196001   1000  591004 591004003     5    20    20
6    591  1960 591196001   1000  591004 591004003     6    20    16
# ... with 3 more variables: NATIVITY <dbl>, EDATTAIN <dbl>,
# EDATTAIND <dbl>
```

# Importing Data Summary

- Recommend tidyverse versions first. If not working try alternatives.

Function	Package	tidyverse	Description
<code>read.csv()</code>	base	No	CSV files
<code>read_csv()</code>	readr	Yes	CSV files
<code>read_excel()</code>	readxl	Yes	Excel files
<code>read.spss()</code>	foreign	No	SPSS files
<code>read_sav()</code>	haven	Yes	SPSS files
<code>read.dta()</code>	foreign	No	Stata files
<code>read.dta13()</code>	readstata13	No	Stata files
<code>read_dta()</code>	haven	Yes	Stata files
<code>read.ssd()</code>	foreign	No	SAS files
<code>read_sas()</code>	haven	Yes	SAS files

# Exporting Data

- CSV are simple and usually preferred.

Function	Package	tidyverse	Description
<code>write.csv()</code>	base	No	CSV files
<code>write_csv()</code>	readr	Yes	CSV files
<code>saveWorkbook()</code>	openxlsx	No	Excel files.
<code>write.foreign()</code>	foreign	No	SPSS, Stata, SAS files
<code>write_sav()</code>	haven	Yes	SPSS files
<code>save.dta13()</code>	readstata13	No	Stata files
<code>write_dta()</code>	haven	Yes	Stata files
<code>write_sas()</code>	haven	Yes	SAS files

# RStudio Data Import Cheatsheet

## Data Import :: CHEAT SHEET

R's **tidyverse** is built around **tidy data** stored in **tibbles**, which are enhanced data frames.



The front side of this sheet shows how to read text files into R with **readr**.



The reverse side shows how to create tibbles with **tibble** and to layout tidy data with **tidyr**.

### OTHER TYPES OF DATA

Try one of the following packages to import other types of files

- **haven** - SPSS, Stata, and SAS files
- **readxl** - excel files (.xls and .xlsx)
- **DBI** - databases
- **jsonlite** - json
- **xmll2** - XML
- **httr** - Web APIs
- **rvest** - HTML (Web Scraping)

## Save Data

Save **x**, an R object, to **path**, a file path, as:

### Comma delimited file

```
write_csv(x, path, na = "NA", append = FALSE,
col_names = !append)
```

### File with arbitrary delimiter

```
write_delim(x, path, delim = " ", na = "NA",
append = FALSE, col_names = !append)
```

### CSV for excel

```
write_excel_csv(x, path, na = "NA", append =
FALSE, col_names = !append)
```

### String to file

```
write_file(x, path, append = FALSE)
```

String vector to file, one element per line

```
write_lines(x,path, na = "NA", append = FALSE)
```

### Object to RDS file

```
write_rds(x, path, compress = c("none", "gz",
"bz2", "xz"), ...)
```

### Tab delimited files

```
write_tsv(x, path, na = "NA", append = FALSE,
col_names = !append)
```

## Read Tabular Data - These functions share the common arguments:

```
read_(file, col_names = TRUE, col_types = NULL, locale = default_locale(), na = c("", "NA"),
quoted_na = TRUE, comment = "", trim_ws = TRUE, skip = 0, n_max = Inf, guess_max = min(1000,
n_max), progress = interactive())
```

### Comma Delimited Files

**read\_csv("file.csv")**

To make file.csv run:  
write\_file(x = "a,b,c\n1,2,3\n4,5,NA", path = "file.csv")

```
a,b,c
1,2,3
4,5,NA
```

### Semi-colon Delimited Files

**read\_csv2("file2.csv")**

write\_file(x = "a;b;c\n1;2;3\n4;5,NA", path = "file2.csv")

```
a;b;c
1;2;3
4;5,NA
```

### Files with Any Delimiter

**read\_delim("file.bt", delim = "|")**

write\_file(x = "a|b|c\n1|2|3\n4|5|NA", path = "file.bt")

```
a|b|c
1|2|3
4|5|NA
```

### Fixed Width Files

**read\_fwf("file.fwf", col\_positions = c(1, 3, 5))**

write\_file(x = "a b c\n1 2 3\n4 5 NA", path = "file.fwf")

```
a b c
1 2 3
4 5 NA
```

### Tab Delimited Files

**read\_tsv("file.tsv")** Also **read\_table()**.

write\_file(x = "\t|b|c\n1|2|3\n4|5|NA", path = "file.tsv")

### USEFUL ARGUMENTS

#### Example file

```
write_file("a,b,c\n1,2,3\n4,5,NA","file.csv")
f <- "file.csv"
```

```
A B C
1 2 3
4 5 NA
```

#### No header

**read\_csv(f, col\_names = FALSE)**

```
x y z
A B C
1 2 3
4 5 NA
```

#### Provide header

**read\_csv(f, col\_names = c("x", "y", "z"))**

#### Skip lines

**read\_csv(f, skip = 1)**

#### Read in a subset

**read\_csv(f, n\_max = 1)**

#### Missing Values

**read\_csv(f, na = c("1", ""))**

## Read Non-Tabular Data

### Read a file into a single string

**read\_file(file, locale = default\_locale())**

### Read each line into its own string

**read\_lines(file, skip = 0, n\_max = -1L, na = character(), locale = default\_locale(), progress = interactive())**

### Read Apache style log files

**read\_log(file, col\_names = FALSE, col\_types = NULL, skip = 0, n\_max = -1, progress = interactive())**

### Read a file into a raw vector

**read\_file\_raw(file)**

### Read each line into a raw vector

**read\_lines\_raw(file, skip = 0, n\_max = -1L, progress = interactive())**

## Data types

readr functions guess the types of each column and convert types when appropriate (but will NOT convert strings to factors automatically).

A message shows the type of each column in the result.

```
## Parsed with column specification:
## cols(
##   age = col_integer(),      age is an integer
##   sex = col_character(),
##   earn = col_double()
## )
```

### 1. Use **problems()** to diagnose problems

```
x <- read_csv("file.csv"); problems(x)
```

### 2. Use a **col\_** function to guide parsing

- **col\_guess()** - the default
- **col\_character()**
- **col\_double()**, **col\_euro\_double()**
- **col\_datetime()**(format = "%") Also **col\_date()**(format = "%"), **col\_time()**(format = "%")
- **col\_factor()**(levels, ordered = FALSE)
- **col\_integer()**
- **col\_logical()**
- **col\_number()**, **col\_numeric()**
- **col\_skip()**

```
x <- read_csv("file.csv", col_types = cols(
A = col_double(),
B = col_logical(),
C = col_factor()))
```

### 3. Else, read in as character vectors then parse with a **parse\_** function.

- **parse\_guess()**
- **parse\_character()**
- **parse\_datetime()** Also **parse\_date()** and **parse\_time()**
- **parse\_double()**
- **parse\_factor()**
- **parse\_integer()**
- **parse\_logical()**
- **parse\_number()**

```
x$A <- parse_number(x$A)
```



# Optional Assignment 3 (assign3.R)

```
##
```

```
## Assignment 3
```

```
##
```

```
# 1. Use an internet search to find some demographic data by region from one country
```

```
# 2. Find a link to download the data in any of the formats covered in the class
# (delimited, Excel, Stata, SPSS, Stata or SAS).
```

```
# Make sure the file is not too large (less than 1MB)
```

```
#
```

```
# If there is no file to download or it is too large go back to step 1 with new search
```

```
# 3. Save your file in the data folder you have been working on in class.
```

```
# 4. Complete the following information below on where you found your data
```

```
# e.g. Data Agency : United Nations Population Division
```

```
# Data Title : Total fertility (TFR)
```

```
# Original File Name: WPP2015_FERT_F04_TOTAL_FERTILITY.xls
```

```
# File Type : Excel
```

```
# URL : https://esa.un.org/unpd/wpp/Download/Standard/Fertility
```

```
#
```

```
# Data Agency :
```

```
# Data Title :
```

```
# Original File Name:
```

```
# File Type :
```