

# intermediate-importing-data

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## Contents

|  |          |
|--|----------|
| <b>Tutorium in R</b>   | <b>1</b> |
| Intermediate Importing data with R . . . . .                   | 1        |
| Initialize MySQL . . . . .                                     | 1        |
| Queries in MySQL . . . . .                                     | 3        |
| HTTP requests . . . . .  | 4        |
| API and JSON . . . . .   | 12       |
| Importing data from other sources (SAS, STATA, SPSS) . . . . . | 15       |

## Tutorium in R

### Intermediate Importing data with R

#### Exercise Intermediate to Importing Data in R - Number 2

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### Initialize MySQL

con is MySQL connection object

```
library(DBI)
perl<- "C:\\Users\\tedoc\\OneDrive\\Dokumente\\R\\win-library\\4.1\\rtools42\\usr\\bin\\perl5.32.1.exe"

con <- dbConnect(RMySQL::MySQL(),
                  dbname = "tweater",
                  host = "courses.csrrinzqubik.us-east-1.rds.amazonaws.com",
                  port = 3306,
                  user = "student",
                  password = "datacamp")

tables <- dbListTables(con)

str(tables) # view all tables
```

```
## chr [1:3] "comments" "tweats" "users"
```

To import user table we use:

```
users <- dbReadTable(con, "users")
print(users)
```

```
## id name login
## 1 1 elisabeth elismith
```

```
## 2 2      mike      mikey
## 3 3      thea      teatime
## 4 4      thomas tomatotom
## 5 5      oliver olivander
## 6 6      kate      katebenn
## 7 7      anjali     lianja
```

To import all tables at once:

```
table_names <- dbListTables(con)
tables <- lapply(table_names, dbReadTable, conn = con)
print(table_names)
```

```
## [1] "comments" "tweats" "users"
```

```
print(tables)
```

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

```
##      id tweet_id user_id      message
## 1  1022      87      7         nice!
## 2  1000      77      7         great!
## 3  1011      49      5         love it
## 4  1012      87      1    awesome! thanks!
## 5  1010      88      6          yuck!
## 6  1026      77      4    not my thing!
## 7  1004      49      1  this is fabulous!
## 8  1030      75      6          so easy!
## 9  1025      88      2          oh yes
## 10 1007      49      3          serious?
## 11 1020      77      1 couldn't be better
## 12 1014      77      1      saved my day
```

```
##
```

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

```
##      id user_id
## 1  75      3
## 2  88      4
## 3  77      6
## 4  87      5
## 5  49      1
## 6  24      7
```

```
##
```

```
##                                     post
## 1                                break egg. bake egg. eat egg.
## 2                                wash strawberries. add ice. blend. enjoy.
## 3                                2 slices of bread. add cheese. grill. heaven.
## 4                                open and crush avocado. add shrimps. perfect starter.
## 5 nachos. add tomato sauce, minced meat and cheese. oven for 10 mins.
## 6                                just eat an apple. simply and healthy.
```

```
##      date
```

```
## 1 2015-09-05
## 2 2015-09-14
## 3 2015-09-21
## 4 2015-09-22
## 5 2015-09-22
## 6 2015-09-24
```

```
##
```

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

```
##   id      name      login
## 1  1 elisabeth  elismith
## 2  2      mike    mikey
## 3  3      thea   teatime
## 4  4    thomas tomatotom
## 5  5    oliver olivander
## 6  6      kate  katebenn
## 7  7    anjali  lianja
```

## Queries in MySQL

Import tweet\_id column of comments where user\_id is 1: elisabeth

```
elisabeth <- dbGetQuery(con, "SELECT tweet_id FROM comments WHERE user_id = 1")
print(elisabeth)
```

```
##   tweet_id
## 1       87
## 2       49
## 3       77
## 4       77
```

Import tweets latest than 2015-09-21

```
latest <- dbGetQuery(con, "SELECT post FROM tweets WHERE date > '2015-09-21'")
print(latest)
```

```
##                                     post
## 1                open and crush avocado. add shrimps. perfect starter.
## 2 nachos. add tomato sauce, minced meat and cheese. oven for 10 mins.
## 3                                just eat an apple. simply and healthy.
```

Select specific information of an object

```
specific <- dbGetQuery(con, "Select message From comments Where tweet_id = 77 AND user_id > 4")
print(specific)
```

```
##   message
## 1  great!
```

Check len of chars with CHAR\_LENGTH()

```
short <- dbGetQuery(con, "Select id, name From users Where CHAR_LENGTH(name)< 5")
print(short)
```

```
##   id name
## 1  2 mike
## 2  3 thea
## 3  6 kate
```

Fetching the data

```
res <- dbSendQuery(con, "SELECT * FROM comments WHERE user_id > 4")
print(res)
```

```
## <MySQLResult:334651504,0,10>
```

```
dbFetch(res, n = 2)
```

```
##   id tweet_id user_id message
## 1 1022      87      7  nice!
```

```
## 2 1000      77      7 great!
```

```
dbFetch(res)
```

```
##      id tweet_id user_id message
```

```
## 1 1011      49      5 love it
```

```
## 2 1010      88      6 yuck!
```

```
## 3 1030      75      6 so easy!
```

```
print(res)
```

```
## <MySQLResult:334651504,0,10>
```

```
#dbClearResult(res)
```

## HTTP requests

Get the data via HTTP GET request

```
library("readr")
```

```
# Import the csv file: pools
```

```
url_csv <- "http://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/swimming_pools."
```

```
pools <- read_csv(url_csv)
```

```
## Rows: 20 Columns: 4
```

```
## -- Column specification -----
```

```
## Delimiter: ","
```

```
## chr (2): Name, Address
```

```
## dbl (2): Latitude, Longitude
```

```
##
```

```
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
# Import the txt file: potatoes
```

```
url_delim <- "http://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/potatoes.txt"
```

```
potatoes <- read_tsv(url_delim)
```

```
## Rows: 160 Columns: 8
```

```
## -- Column specification -----
```

```
## Delimiter: "\t"
```

```
## dbl (8): area, temp, size, storage, method, texture, flavor, moistness
```

```
##
```

```
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
print(pools)
```

```
## # A tibble: 20 x 4
```

```
##   Name                Address                Latitude Longitude
```

```
##   <chr>                <chr>                <dbl>    <dbl>
```

```
## 1 Acacia Ridge Leisure Centre 1391 Beaudesert~ -27.6    153.
```

```
## 2 Bellbowrie Pool          Sugarwood Stree~ -27.6    153.
```

```
## 3 Carole Park              Cnr Boundary Ro~ -27.6    153.
```

```
## 4 Centenary Pool (inner City) 400 Gregory Ter~ -27.5    153.
```

```
## 5 Chermside Pool           375 Hamilton Ro~ -27.4    153.
```

```
## 6 Colmslie Pool (Morningside) 400 Lytton Road~ -27.5    153.
```

```
## 7 Spring Hill Baths (inner City) 14 Torrington S~ -27.5    153.
```

```
## 8 Dunlop Park Pool (Corinda) 794 Oxley Road,~ -27.5 153.
## 9 Fortitude Valley Pool 432 Wickham Str~ -27.5 153.
## 10 Hibiscus Sports Complex (upper MtGravatt) 90 Klumpp Road,~ -27.6 153.
## 11 Ithaca Pool ( Paddington) 131 Caxton Stre~ -27.5 153.
## 12 Jindalee Pool 11 Yallambee Ro~ -27.5 153.
## 13 Manly Pool 1 Fairlead Cres~ -27.5 153.
## 14 Mt Gravatt East Aquatic Centre Cnr wecker Road~ -27.5 153.
## 15 Musgrave Park Pool (South Brisbane) 100 Edmonstone ~ -27.5 153.
## 16 Newmarket Pool 71 Alderson Str~ -27.4 153.
## 17 Runcorn Pool 37 Bonemill Roa~ -27.6 153.
## 18 Sandgate Pool 231 Flinders Pa~ -27.3 153.
## 19 Langlands Parks Pool (Stones Corner) 5 Panitya Stree~ -27.5 153.
## 20 Yeronga Park Pool 81 School Road,~ -27.5 153.
```

```
print(potatoes)
```

```
## # A tibble: 160 x 8
##   area temp size storage method texture flavor moistness
##   <dbl> <dbl> <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1     1     1     1     1     1     1     2.9     3.2     3
## 2     1     1     1     1     2     2     2.3     2.5     2.6
## 3     1     1     1     1     3     3     2.5     2.8     2.8
## 4     1     1     1     1     4     4     2.1     2.9     2.4
## 5     1     1     1     1     5     5     1.9     2.8     2.2
## 6     1     1     1     2     1     1     1.8     3     1.7
## 7     1     1     1     2     2     2     2.6     3.1     2.4
## 8     1     1     1     2     3     3     3     3     2.9
## 9     1     1     1     2     4     4     2.2     3.2     2.5
## 10    1     1     1     2     5     5     2     2.8     1.9
## # ... with 150 more rows
```

For secure connection to **https** we use **read.csv()**

To download data use:

```
library(readxl)
library(gdata)
```

```
## gdata: Unable to locate valid perl interpreter
## gdata:
## gdata: read.xls() will be unable to read Excel XLS and XLSX files
## gdata: unless the 'perl=' argument is used to specify the location of a
## gdata: valid perl intrpreter.
## gdata:
## gdata: (To avoid display of this message in the future, please ensure
## gdata: perl is installed and available on the executable search path.)
## gdata: Unable to load perl libraries needed by read.xls()
## gdata: to support 'XLX' (Excel 97-2004) files.
##
## gdata: Unable to load perl libraries needed by read.xls()
## gdata: to support 'XLSX' (Excel 2007+) files.
##
## gdata: Run the function 'installXLSXsupport()'
## gdata: to automatically download and install the perl
```

```
## gdata: libraries needed to support Excel XLS and XLSX formats.
##
## Attaching package: 'gdata'
## The following object is masked from 'package:stats':
##
##     nobs
## The following object is masked from 'package:utils':
##
##     object.size
## The following object is masked from 'package:base':
##
##     startsWith
url_xls <- "http://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/latitude.xls"
excel_gdata <- read.xls(url_xls, perl = perl)

## Warning in system(cmd, intern = intern, wait = wait | intern,
## show.output.on.console = wait, : running command 'C:\Windows\system32\cmd.exe /c
## ftype perl' had status 2

## Warning in system(cmd, intern = intern, wait = wait | intern,
## show.output.on.console = wait, : running command 'C:\Windows\system32\cmd.exe /c
## ftype perl' had status 2
print(excel_gdata)

##           country      X1700
## 1      Afghanistan 34.5650000
## 2 Akrotiri and Dhekelia 34.6166667
## 3      Albania 41.3120000
## 4      Algeria 36.7200000
## 5 American Samoa -14.3070000
## 6      Andorra 42.5460000
## 7      Angola -8.8430000
## 8      Anguilla 18.2500000
## 9 Antigua and Barbuda 17.0720000
## 10     Argentina -36.6760000
## 11     Armenia 40.2540000
## 12     Aruba 12.5130000
## 13     Australia -32.2190000
## 14     Austria 48.2310000
## 15 Azerbaijan 40.3520000
## 16     Bahamas 24.7000000
## 17     Bahrain 26.0240000
## 18 Bangladesh 23.8800000
## 19 Barbados 13.1790000
## 20     Belarus 53.5470900
## 21     Belgium 50.8370000
## 22     Belize 17.8430000
## 23     Benin 6.3640000
## 24     Bermuda 32.2170000
## 25     Bhutan 27.4790000
## 26     Bolivia -15.1900000
```

|       |                             |             |
|-------|-----------------------------|-------------|
| ## 27 | Bosnia and Herzegovina      | 44.1750100  |
| ## 28 | Botswana                    | -21.5360000 |
| ## 29 | Brazil                      | -19.5570000 |
| ## 30 | British Virgin Islands      | 18.5000000  |
| ## 31 | Brunei                      | 4.5010000   |
| ## 32 | Bulgaria                    | 42.0730000  |
| ## 33 | Burkina Faso                | 12.0490000  |
| ## 34 | Burundi                     | -3.3650000  |
| ## 35 | Cambodia                    | 12.0260000  |
| ## 36 | Cameroon                    | 10.7300000  |
| ## 37 | Canada                      | 43.7270000  |
| ## 38 | Cape Verde                  | 15.0910000  |
| ## 39 | Cayman Islands              | 19.3190000  |
| ## 40 | Central African Rep.        | 4.3310000   |
| ## 41 | Chad                        | 10.3770000  |
| ## 42 | Channel Islands             | 49.2170000  |
| ## 43 | Chile                       | -33.5540000 |
| ## 44 | China                       | 29.5610000  |
| ## 45 | Christmas Island            | -10.5000000 |
| ## 46 | Cocos Island                | -12.5000000 |
| ## 47 | Colombia                    | 4.7880000   |
| ## 48 | Comoros                     | -11.6710000 |
| ## 49 | Congo, Dem. Rep.            | -2.9202760  |
| ## 50 | Congo, Rep.                 | -3.6840000  |
| ## 51 | Cook Islands                | -21.2333333 |
| ## 52 | Costa Rica                  | 9.9410000   |
| ## 53 | Cote d'Ivoire               | 5.4960000   |
| ## 54 | Croatia                     | 45.1100800  |
| ## 55 | Cuba                        | 23.0840000  |
| ## 56 | Cyprus                      | 35.0810000  |
| ## 57 | Czech Rep.                  | 49.7792900  |
| ## 58 | Denmark                     | 55.7180000  |
| ## 59 | Djibouti                    | 11.5050000  |
| ## 60 | Dominica                    | 15.4330000  |
| ## 61 | Dominican Rep.              | 18.5610000  |
| ## 62 | East Germany                | NA          |
| ## 63 | Ecuador                     | -2.0620000  |
| ## 64 | Egypt                       | 29.9960000  |
| ## 65 | El Salvador                 | 13.7750000  |
| ## 66 | Equatorial Guinea           | 2.3260000   |
| ## 67 | Eritrea                     | 15.3120000  |
| ## 68 | Estonia                     | 58.6850000  |
| ## 69 | Ethiopia                    | 9.0070000   |
| ## 70 | Faeroe Islands              | 62.0000000  |
| ## 71 | Falkland Islands (Malvinas) | -51.7500000 |
| ## 72 | Fiji                        | -17.8270000 |
| ## 73 | Finland                     | 60.2120000  |
| ## 74 | France                      | 48.8570000  |
| ## 75 | French Guiana               | 3.9880000   |
| ## 76 | French Polynesia            | -17.6660000 |
| ## 77 | Gabon                       | 0.3720000   |
| ## 78 | Gambia                      | 13.2570000  |
| ## 79 | Georgia                     | 42.0340000  |
| ## 80 | Germany                     | 48.1610000  |

|        |                  |             |
|--------|------------------|-------------|
| ## 81  | Ghana            | 6.6940000   |
| ## 82  | Gibraltar        | 36.1333333  |
| ## 83  | Greece           | 38.0580000  |
| ## 84  | Greenland        | 64.2170000  |
| ## 85  | Grenada          | 12.1150000  |
| ## 86  | Guadeloupe       | 16.1630000  |
| ## 87  | Guam             | 13.4400000  |
| ## 88  | Guatemala        | 14.6220000  |
| ## 89  | Guernsey         | 49.4666667  |
| ## 90  | Guinea           | 11.6710000  |
| ## 91  | Guinea-Bissau    | 12.2620000  |
| ## 92  | Guyana           | 5.7610000   |
| ## 93  | Haiti            | 18.9320000  |
| ## 94  | Holy See         | 41.9000000  |
| ## 95  | Honduras         | 14.1940000  |
| ## 96  | Hong Kong, China | 22.7040000  |
| ## 97  | Hungary          | 47.4190000  |
| ## 98  | Iceland          | 63.8920000  |
| ## 99  | India            | 25.2740000  |
| ## 100 | Indonesia        | -6.5620000  |
| ## 101 | Iran             | 35.3800000  |
| ## 102 | Iraq             | 33.3140000  |
| ## 103 | Ireland          | 54.6110000  |
| ## 104 | Isle of Man      | 54.2250000  |
| ## 105 | Israel           | 32.0840000  |
| ## 106 | Italy            | 45.4150000  |
| ## 107 | Jamaica          | 18.0550000  |
| ## 108 | Japan            | 35.7120000  |
| ## 109 | Jersey           | 49.2500000  |
| ## 110 | Jordan           | 31.6020000  |
| ## 111 | Kazakhstan       | 44.3080000  |
| ## 112 | Kenya            | -0.5130000  |
| ## 113 | Kiribati         | 1.8470000   |
| ## 114 | Korea, Dem. Rep. | 39.5270000  |
| ## 115 | Korea, Rep.      | 37.5530000  |
| ## 116 | Kosovo           | 42.5833333  |
| ## 117 | Kuwait           | 29.3260000  |
| ## 118 | Kyrgyzstan       | 41.4894000  |
| ## 119 | Laos             | 16.5470000  |
| ## 120 | Latvia           | 56.8580000  |
| ## 121 | Lebanon          | 34.1110000  |
| ## 122 | Lesotho          | -29.5950000 |
| ## 123 | Liberia          | 6.3850000   |
| ## 124 | Libya            | 32.6070000  |
| ## 125 | Liechtenstein    | 47.1410000  |
| ## 126 | Lithuania        | 55.3180000  |
| ## 127 | Luxembourg       | 49.7800000  |
| ## 128 | Macao, China     | 22.5060000  |
| ## 129 | Macedonia, FYR   | 41.5738100  |
| ## 130 | Madagascar       | -18.9580000 |
| ## 131 | Malawi           | -15.8110000 |
| ## 132 | Malaysia         | 3.2690000   |
| ## 133 | Maldives         | 3.2500000   |
| ## 134 | Mali             | 12.5080000  |



|        |                                  |             |
|--------|----------------------------------|-------------|
| ## 135 | Malta                            | 35.8870000  |
| ## 136 | Marshall Islands                 | 9.0000000   |
| ## 137 | Martinique                       | 14.6540000  |
| ## 138 | Mauritania                       | 17.9250000  |
| ## 139 | Mauritius                        | -20.2320000 |
| ## 140 | Mayotte                          | -12.8333333 |
| ## 141 | Mexico                           | 16.7590000  |
| ## 142 | Micronesia, Fed. Sts.            | 7.3570000   |
| ## 143 | Moldova                          | 47.1660000  |
| ## 144 | Monaco                           | 43.7460000  |
| ## 145 | Mongolia                         | 47.4930000  |
| ## 146 | Montenegro                       | 42.7889900  |
| ## 147 | Montserrat                       | 16.7500000  |
| ## 148 | Morocco                          | 33.5930000  |
| ## 149 | Mozambique                       | -18.4990000 |
| ## 150 | Myanmar                          | 17.6790000  |
| ## 151 | Namibia                          | -17.9790000 |
| ## 152 | Nauru                            | -0.5333333  |
| ## 153 | Nepal                            | 27.7120000  |
| ## 154 | Netherlands                      | 51.8740000  |
| ## 155 | Netherlands Antilles             | 12.1920000  |
| ## 156 | New Caledonia                    | -21.3290000 |
| ## 157 | New Zealand                      | -36.8920000 |
| ## 158 | Nicaragua                        | 12.2110000  |
| ## 159 | Niger                            | 13.8760000  |
| ## 160 | Nigeria                          | 6.5430000   |
| ## 161 | Niue                             | -19.0333333 |
| ## 162 | Norfolk Island                   | -29.0333333 |
| ## 163 | Northern Mariana Islands         | 15.1780000  |
| ## 164 | Norway                           | 59.9770000  |
| ## 165 | Oman                             | 20.4450000  |
| ## 166 | Pakistan                         | 31.1730000  |
| ## 167 | Palau                            | 7.5000000   |
| ## 168 | Panama                           | 9.2060000   |
| ## 169 | Papua New Guinea                 | -6.6000000  |
| ## 170 | Paraguay                         | -25.5830000 |
| ## 171 | Peru                             | -11.7940000 |
| ## 172 | Philippines                      | 13.9220000  |
| ## 173 | Pitcairn                         | -25.0666667 |
| ## 174 | Poland                           | 50.2440000  |
| ## 175 | Portugal                         | 38.8160000  |
| ## 176 | Puerto Rico                      | 18.2250000  |
| ## 177 | Qatar                            | 25.3090000  |
| ## 178 | Reunion                          | -20.9540000 |
| ## 179 | Romania                          | 44.5260000  |
| ## 180 | Russia                           | 55.6750000  |
| ## 181 | Rwanda                           | -2.0320000  |
| ## 182 | Saint Barthélemy                 | 17.9000000  |
| ## 183 | Saint Helena                     | -15.9500000 |
| ## 184 | Saint Kitts and Nevis            | 17.3270000  |
| ## 185 | Saint Lucia                      | 13.8980000  |
| ## 186 | Saint Martin                     | 18.0833333  |
| ## 187 | Saint Vincent and the Grenadines | 13.2540000  |
| ## 188 | Saint-Pierre-et-Miquelon         | 46.8333333  |

|        |                          |             |
|--------|--------------------------|-------------|
| ## 189 | Samoa                    | -13.6330000 |
| ## 190 | San Marino               | 43.7666667  |
| ## 191 | Sao Tome and Principe    | 1.0000000   |
| ## 192 | Saudi Arabia             | 23.0690000  |
| ## 193 | Senegal                  | 14.7720000  |
| ## 194 | Serbia                   | 44.0467300  |
| ## 195 | Seychelles               | -4.6640000  |
| ## 196 | Sierra Leone             | 8.7010000   |
| ## 197 | Singapore                | 1.3550000   |
| ## 198 | Slovak Republic          | 48.7853800  |
| ## 199 | Slovenia                 | 46.1279000  |
| ## 200 | Solomon Islands          | -9.6250000  |
| ## 201 | Somalia                  | 10.6330000  |
| ## 202 | South Africa             | -29.1300000 |
| ## 203 | Spain                    | 37.3980000  |
| ## 204 | Sri Lanka                | 6.8680000   |
| ## 205 | Sudan                    | 14.0430000  |
| ## 206 | Suriname                 | 5.6050000   |
| ## 207 | Svalbard                 | 78.0000000  |
| ## 208 | Swaziland                | -26.5450000 |
| ## 209 | Sweden                   | 59.2780000  |
| ## 210 | Switzerland              | 47.4080000  |
| ## 211 | Syria                    | 33.4580000  |
| ## 212 | Taiwan                   | 23.6448500  |
| ## 213 | Tajikistan               | 37.8060000  |
| ## 214 | Tanzania                 | -2.1540000  |
| ## 215 | Thailand                 | 13.7700000  |
| ## 216 | Timor-Leste              | -8.8333333  |
| ## 217 | Togo                     | 6.1940000   |
| ## 218 | Tokelau                  | -9.0000000  |
| ## 219 | Tonga                    | -21.1730000 |
| ## 220 | Trinidad and Tobago      | 10.4180000  |
| ## 221 | Tunisia                  | 36.8160000  |
| ## 222 | Turkey                   | 41.2020000  |
| ## 223 | Turkmenistan             | 39.1293800  |
| ## 224 | Turks and Caicos Islands | 21.7500000  |
| ## 225 | Tuvalu                   | -8.0000000  |
| ## 226 | Uganda                   | 0.2280000   |
| ## 227 | Ukraine                  | 50.2810000  |
| ## 228 | United Arab Emirates     | 23.3900000  |
| ## 229 | United Kingdom           | 51.5100000  |
| ## 230 | United States            | 34.3600000  |
| ## 231 | Uruguay                  | -34.8220000 |
| ## 232 | USSR                     | NA          |
| ## 233 | Uzbekistan               | 41.2720000  |
| ## 234 | Wallis et Futuna         | -13.3000000 |
| ## 235 | Vanuatu                  | -15.2330000 |
| ## 236 | Venezuela                | 9.8430000   |
| ## 237 | West Bank and Gaza       | 31.4166667  |
| ## 238 | West Germany             | NA          |
| ## 239 | Western Sahara           | 24.6191300  |
| ## 240 | Vietnam                  | 10.7980000  |
| ## 241 | Virgin Islands (U.S.)    | 17.7360000  |
| ## 242 | Yemen, Rep.              | 15.2280000  |

```
## 243                Yugoslavia      NA
## 244                Zambia -12.9420000
## 245                Zimbabwe -17.8760000
## 246                Åland  60.0000000
```

To workaround reading excel file with read\_excel() we need to:

```
download.file(url_xls, destfile = "local_latitude.xls") #download file locally

#excel_readxl <- read_excel("local_latitude.xls") #read it
```

Download file and read it as RData file

```
url_rdata <- "https://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/wine.RData"
download.file(url_rdata, destfile = "wine_local.RData")
load("wine_local.RData")

summary(wine)
```

```
##      Alcohol      Malic acid      Ash      Alkalinity of ash
##  Min.   :11.03  Min.   :0.74  Min.   :1.360  Min.   :10.60
## 1st Qu.:12.36  1st Qu.:1.60  1st Qu.:2.210  1st Qu.:17.20
## Median :13.05  Median :1.87  Median :2.360  Median :19.50
## Mean   :12.99  Mean   :2.34  Mean   :2.366  Mean   :19.52
## 3rd Qu.:13.67  3rd Qu.:3.10  3rd Qu.:2.560  3rd Qu.:21.50
## Max.   :14.83  Max.   :5.80  Max.   :3.230  Max.   :30.00
##  Magnesium    Total phenols    Flavanoids    Nonflavanoid phenols
##  Min.   : 70.00  Min.   :0.980  Min.   :0.340  Min.   :0.1300
## 1st Qu.: 88.00  1st Qu.:1.740  1st Qu.:1.200  1st Qu.:0.2700
## Median : 98.00  Median :2.350  Median :2.130  Median :0.3400
## Mean   : 99.59  Mean   :2.292  Mean   :2.023  Mean   :0.3623
## 3rd Qu.:107.00  3rd Qu.:2.800  3rd Qu.:2.860  3rd Qu.:0.4400
## Max.   :162.00  Max.   :3.880  Max.   :5.080  Max.   :0.6600
## Proanthocyanins Color intensity    Hue      Proline
##  Min.   :0.410  Min.   : 1.280  Min.   :1.270  Min.   : 278.0
## 1st Qu.:1.250  1st Qu.: 3.210  1st Qu.:1.930  1st Qu.: 500.0
## Median :1.550  Median : 4.680  Median :2.780  Median : 672.0
## Mean   :1.587  Mean   : 5.055  Mean   :2.604  Mean   : 745.1
## 3rd Qu.:1.950  3rd Qu.: 6.200  3rd Qu.:3.170  3rd Qu.: 985.0
## Max.   :3.580  Max.   :13.000  Max.   :4.000  Max.   :1680.0
```

Get content of HTTP request

```
library("httr")
url <- "http://www.example.com/"
resp <- GET(url)

print(resp)
```

```
## Response [http://www.example.com/]
##   Date: 2022-03-26 11:44
##   Status: 200
##   Content-Type: text/html; charset=UTF-8
##   Size: 1.26 kB
## <!doctype html>
## <html>
## <head>
```

```
## <title>Example Domain</title>
##
## <meta charset="utf-8" />
## <meta http-equiv="Content-type" content="text/html; charset=utf-8" />
## <meta name="viewport" content="width=device-width, initial-scale=1" />
## <style type="text/css">
## body {
## ...
```

```
raw_content <- content(as = "raw", resp)
head(raw_content)
```

```
## [1] 3c 21 64 6f 63 74
```

## API and JSON

Read JSON data

```
library("jsonlite")
wine_json <- '{"name":"Chateau Migraine", "year":1997, "alcohol_pct":12.4, "color":"red", "awarded":fal...'

wine <- fromJSON(wine_json)
str(wine)
```

```
## List of 5
## $ name      : chr "Chateau Migraine"
## $ year      : int 1997
## $ alcohol_pct: num 12.4
## $ color     : chr "red"
## $ awarded   : logi FALSE
```

```
url_sw4 <- "http://www.omdbapi.com/?apikey=72bc447a&i=tt0076759&r=json"
url_sw3 <- "http://www.omdbapi.com/?apikey=72bc447a&i=tt0121766&r=json"
```

```
sw4 <- fromJSON(url_sw4)
sw3 <- fromJSON(url_sw3)
```

```
# Print out the Title element of both lists
sw4$Title
```

```
## [1] "Star Wars"
```

```
sw3$Title
```

```
## [1] "Star Wars: Episode III - Revenge of the Sith"
```

```
# Is the release year of sw4 later than sw3?
sw4$Year > sw3$Year
```

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

```
print(sw4)
```

```
## $Title
## [1] "Star Wars"
##
## $Year
## [1] "1977"
##
```

```

## $Rated
## [1] "PG"
##
## $Released
## [1] "25 May 1977"
##
## $Runtime
## [1] "121 min"
##
## $Genre
## [1] "Action, Adventure, Fantasy"
##
## $Director
## [1] "George Lucas"
##
## $Writer
## [1] "George Lucas"
##
## $Actors
## [1] "Mark Hamill, Harrison Ford, Carrie Fisher"
##
## $Plot
## [1] "Luke Skywalker joins forces with a Jedi Knight, a cocky pilot, a Wookiee and two droids to save
##
## $Language
## [1] "English"
##
## $Country
## [1] "United States"
##
## $Awards
## [1] "Won 7 Oscars. 63 wins & 29 nominations total"
##
## $Poster
## [1] "https://m.media-amazon.com/images/M/MV5BNzVlY2MwMjktM2E4OS00Y2Y3LWE3ZjctYzhkZGM3YzA1ZWY2XkEyXkF
##
## $Ratings
##           Source Value
## 1 Internet Movie Database 8.6/10
## 2       Rotten Tomatoes   92%
## 3       Metacritic 90/100
##
## $Metascore
## [1] "90"
##
## $imdbRating
## [1] "8.6"
##
## $imdbVotes
## [1] "1,312,386"
##
## $imdbID
## [1] "tt0076759"
##

```

```
## $Type
## [1] "movie"
##
## $DVD
## [1] "06 Dec 2005"
##
## $BoxOffice
## [1] "$460,998,507"
##
## $Production
## [1] "N/A"
##
## $Website
## [1] "N/A"
##
## $Response
## [1] "True"
```

Build from Json object

```
# Challenge 1
json1 <- '[1, 2, 3, 4, 5, 6]'
fromJSON(json1)
```

```
## [1] 1 2 3 4 5 6
```

```
# Challenge 2
json2 <- '{"a": [1, 2, 3], "b": [4, 5, 6]}'
fromJSON(json2)
```

```
## $a
## [1] 1 2 3
##
## $b
## [1] 4 5 6
```

```
# Challenge 1
json1 <- '[[1, 2], [3, 4]]'
fromJSON(json1)
```

```
##      [,1] [,2]
## [1,]    1    2
## [2,]    3    4
```

```
# Challenge 2
json2 <- ' [{"a": 1, "b": 2}, {"a": 3, "b": 4}, {"a": 5, "b": 6}] '
fromJSON(json2)
```

```
##   a b
## 1 1 2
## 2 3 4
## 3 5 6
```

Convert the data to JSON

```
url_csv <- "http://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/water.csv"

# Import the .csv file located at url_csv
water <- read.csv(url_csv, stringsAsFactors=FALSE)
```

```
# Convert the data file according to the requirements
water_json <- toJSON(water)
print(water_json)
```

```
## [{"water": "Algeria", "X1992": 0.064, "X2002": 0.017}, {"water": "American Samoa"}, {"water": "Angola", "X1992": 0.017, "X2002": 0.017}, {"water": "Antigua and Barbuda", "X1992": 0.017, "X2002": 0.017}, {"water": "Argentina", "X1992": 0.017, "X2002": 0.017}, {"water": "Armenia", "X1992": 0.017, "X2002": 0.017}, {"water": "Australia", "X1992": 0.017, "X2002": 0.017}, {"water": "Austria", "X1992": 0.017, "X2002": 0.017}, {"water": "Azerbaijan", "X1992": 0.017, "X2002": 0.017}, {"water": "Bahamas", "X1992": 0.017, "X2002": 0.017}, {"water": "Bahrain", "X1992": 0.017, "X2002": 0.017}, {"water": "Bangladesh", "X1992": 0.017, "X2002": 0.017}, {"water": "Barbados", "X1992": 0.017, "X2002": 0.017}, {"water": "Belarus", "X1992": 0.017, "X2002": 0.017}, {"water": "Belgium", "X1992": 0.017, "X2002": 0.017}, {"water": "Belize", "X1992": 0.017, "X2002": 0.017}, {"water": "Benin", "X1992": 0.017, "X2002": 0.017}, {"water": "Bhutan", "X1992": 0.017, "X2002": 0.017}, {"water": "Bolivia", "X1992": 0.017, "X2002": 0.017}, {"water": "Bosnia and Herzegovina", "X1992": 0.017, "X2002": 0.017}, {"water": "Botswana", "X1992": 0.017, "X2002": 0.017}, {"water": "Brazil", "X1992": 0.017, "X2002": 0.017}, {"water": "Brunei Darussalam", "X1992": 0.017, "X2002": 0.017}, {"water": "Bulgaria", "X1992": 0.017, "X2002": 0.017}, {"water": "Burkina Faso", "X1992": 0.017, "X2002": 0.017}, {"water": "Burundi", "X1992": 0.017, "X2002": 0.017}, {"water": "Cambodia", "X1992": 0.017, "X2002": 0.017}, {"water": "Cameroon", "X1992": 0.017, "X2002": 0.017}, {"water": "Canada", "X1992": 0.017, "X2002": 0.017}, {"water": "Cape Verde", "X1992": 0.017, "X2002": 0.017}, {"water": "Cayman Islands", "X1992": 0.017, "X2002": 0.017}, {"water": "Central African Republic", "X1992": 0.017, "X2002": 0.017}, {"water": "Chad", "X1992": 0.017, "X2002": 0.017}, {"water": "Chile", "X1992": 0.017, "X2002": 0.017}, {"water": "China", "X1992": 0.017, "X2002": 0.017}, {"water": "Colombia", "X1992": 0.017, "X2002": 0.017}, {"water": "Comoros", "X1992": 0.017, "X2002": 0.017}, {"water": "Congo", "X1992": 0.017, "X2002": 0.017}, {"water": "Congo (Kinshasa)", "X1992": 0.017, "X2002": 0.017}, {"water": "Costa Rica", "X1992": 0.017, "X2002": 0.017}, {"water": "Cote d'Ivoire", "X1992": 0.017, "X2002": 0.017}, {"water": "Croatia", "X1992": 0.017, "X2002": 0.017}, {"water": "Cuba", "X1992": 0.017, "X2002": 0.017}, {"water": "Cyprus", "X1992": 0.017, "X2002": 0.017}, {"water": "Czechia", "X1992": 0.017, "X2002": 0.017}, {"water": "Democratic Republic of the Congo", "X1992": 0.017, "X2002": 0.017}, {"water": "Denmark", "X1992": 0.017, "X2002": 0.017}, {"water": "Djibouti", "X1992": 0.017, "X2002": 0.017}, {"water": "Dominica", "X1992": 0.017, "X2002": 0.017}, {"water": "Dominican Republic", "X1992": 0.017, "X2002": 0.017}, {"water": "East Asia", "X1992": 0.017, "X2002": 0.017}, {"water": "Ecuador", "X1992": 0.017, "X2002": 0.017}, {"water": "Egypt", "X1992": 0.017, "X2002": 0.017}, {"water": "El Salvador", "X1992": 0.017, "X2002": 0.017}, {"water": "Equatorial Guinea", "X1992": 0.017, "X2002": 0.017}, {"water": "Eritrea", "X1992": 0.017, "X2002": 0.017}, {"water": "Estonia", "X1992": 0.017, "X2002": 0.017}, {"water": "Ethiopia", "X1992": 0.017, "X2002": 0.017}, {"water": "European Union", "X1992": 0.017, "X2002": 0.017}, {"water": "Fiji", "X1992": 0.017, "X2002": 0.017}, {"water": "Finland", "X1992": 0.017, "X2002": 0.017}, {"water": "France", "X1992": 0.017, "X2002": 0.017}, {"water": "Gabon", "X1992": 0.017, "X2002": 0.017}, {"water": "Gambia", "X1992": 0.017, "X2002": 0.017}, {"water": "Georgia", "X1992": 0.017, "X2002": 0.017}, {"water": "Germany", "X1992": 0.017, "X2002": 0.017}, {"water": "Ghana", "X1992": 0.017, "X2002": 0.017}, {"water": "Greece", "X1992": 0.017, "X2002": 0.017}, {"water": "Guatemala", "X1992": 0.017, "X2002": 0.017}, {"water": "Guinea", "X1992": 0.017, "X2002": 0.017}, {"water": "Guinea-Bissau", "X1992": 0.017, "X2002": 0.017}, {"water": "Guyana", "X1992": 0.017, "X2002": 0.017}, {"water": "Haiti", "X1992": 0.017, "X2002": 0.017}, {"water": "Honduras", "X1992": 0.017, "X2002": 0.017}, {"water": "Hungary", "X1992": 0.017, "X2002": 0.017}, {"water": "Iceland", "X1992": 0.017, "X2002": 0.017}, {"water": "India", "X1992": 0.017, "X2002": 0.017}, {"water": "Indonesia", "X1992": 0.017, "X2002": 0.017}, {"water": "Iran", "X1992": 0.017, "X2002": 0.017}, {"water": "Iraq", "X1992": 0.017, "X2002": 0.017}, {"water": "Israel", "X1992": 0.017, "X2002": 0.017}, {"water": "Italy", "X1992": 0.017, "X2002": 0.017}, {"water": "Jamaica", "X1992": 0.017, "X2002": 0.017}, {"water": "Japan", "X1992": 0.017, "X2002": 0.017}, {"water": "Jordan", "X1992": 0.017, "X2002": 0.017}, {"water": "Kazakhstan", "X1992": 0.017, "X2002": 0.017}, {"water": "Kenya", "X1992": 0.017, "X2002": 0.017}, {"water": "Korea", "X1992": 0.017, "X2002": 0.017}, {"water": "Korea (South)", "X1992": 0.017, "X2002": 0.017}, {"water": "Korea (North)", "X1992": 0.017, "X2002": 0.017}, {"water": "Kuwait", "X1992": 0.017, "X2002": 0.017}, {"water": "Kyrgyzstan", "X1992": 0.017, "X2002": 0.017}, {"water": "Laos", "X1992": 0.017, "X2002": 0.017}, {"water": "Latvia", "X1992": 0.017, "X2002": 0.017}, {"water": "Lebanon", "X1992": 0.017, "X2002": 0.017}, {"water": "Lesotho", "X1992": 0.017, "X2002": 0.017}, {"water": "Liberia", "X1992": 0.017, "X2002": 0.017}, {"water": "Liechtenstein", "X1992": 0.017, "X2002": 0.017}, {"water": "Lithuania", "X1992": 0.017, "X2002": 0.017}, {"water": "Luxembourg", "X1992": 0.017, "X2002": 0.017}, {"water": "Madagascar", "X1992": 0.017, "X2002": 0.017}, {"water": "Malawi", "X1992": 0.017, "X2002": 0.017}, {"water": "Malaysia", "X1992": 0.017, "X2002": 0.017}, {"water": "Maldives", "X1992": 0.017, "X2002": 0.017}, {"water": "Mali", "X1992": 0.017, "X2002": 0.017}, {"water": "Malta", "X1992": 0.017, "X2002": 0.017}, {"water": "Mauritania", "X1992": 0.017, "X2002": 0.017}, {"water": "Mauritius", "X1992": 0.017, "X2002": 0.017}, {"water": "Mexico", "X1992": 0.017, "X2002": 0.017}, {"water": "Moldova", "X1992": 0.017, "X2002": 0.017}, {"water": "Monaco", "X1992": 0.017, "X2002": 0.017}, {"water": "Mongolia", "X1992": 0.017, "X2002": 0.017}, {"water": "Montenegro", "X1992": 0.017, "X2002": 0.017}, {"water": "Morocco", "X1992": 0.017, "X2002": 0.017}, {"water": "Mozambique", "X1992": 0.017, "X2002": 0.017}, {"water": "Myanmar", "X1992": 0.017, "X2002": 0.017}, {"water": "Namibia", "X1992": 0.017, "X2002": 0.017}, {"water": "Nauru", "X1992": 0.017, "X2002": 0.017}, {"water": "Nepal", "X1992": 0.017, "X2002": 0.017}, {"water": "Netherlands", "X1992": 0.017, "X2002": 0.017}, {"water": "New Zealand", "X1992": 0.017, "X2002": 0.017}, {"water": "Nicaragua", "X1992": 0.017, "X2002": 0.017}, {"water": "Niger", "X1992": 0.017, "X2002": 0.017}, {"water": "Nigeria", "X1992": 0.017, "X2002": 0.017}, {"water": "North Macedonia", "X1992": 0.017, "X2002": 0.017}, {"water": "Norway", "X1992": 0.017, "X2002": 0.017}, {"water": "Oman", "X1992": 0.017, "X2002": 0.017}, {"water": "Pakistan", "X1992": 0.017, "X2002": 0.017}, {"water": "Palestine", "X1992": 0.017, "X2002": 0.017}, {"water": "Panama", "X1992": 0.017, "X2002": 0.017}, {"water": "Papua New Guinea", "X1992": 0.017, "X2002": 0.017}, {"water": "Paraguay", "X1992": 0.017, "X2002": 0.017}, {"water": "Peru", "X1992": 0.017, "X2002": 0.017}, {"water": "Philippines", "X1992": 0.017, "X2002": 0.017}, {"water": "Poland", "X1992": 0.017, "X2002": 0.017}, {"water": "Portugal", "X1992": 0.017, "X2002": 0.017}, {"water": "Puerto Rico", "X1992": 0.017, "X2002": 0.017}, {"water": "Qatar", "X1992": 0.017, "X2002": 0.017}, {"water": "Romania", "X1992": 0.017, "X2002": 0.017}, {"water": "Rwanda", "X1992": 0.017, "X2002": 0.017}, {"water": "Russia", "X1992": 0.017, "X2002": 0.017}, {"water": "Rwanda", "X1992": 0.017, "X2002": 0.017}, {"water": "Saudi Arabia", "X1992": 0.017, "X2002": 0.017}, {"water": "Senegal", "X1992": 0.017, "X2002": 0.017}, {"water": "Serbia", "X1992": 0.017, "X2002": 0.017}, {"water": "Sierra Leone", "X1992": 0.017, "X2002": 0.017}, {"water": "Singapore", "X1992": 0.017, "X2002": 0.017}, {"water": "Slovakia", "X1992": 0.017, "X2002": 0.017}, {"water": "Slovenia", "X1992": 0.017, "X2002": 0.017}, {"water": "Somalia", "X1992": 0.017, "X2002": 0.017}, {"water": "South Africa", "X1992": 0.017, "X2002": 0.017}, {"water": "South Asia", "X1992": 0.017, "X2002": 0.017}, {"water": "South Korea", "X1992": 0.017, "X2002": 0.017}, {"water": "South Sudan", "X1992": 0.017, "X2002": 0.017}, {"water": "Spain", "X1992": 0.017, "X2002": 0.017}, {"water": "Sri Lanka", "X1992": 0.017, "X2002": 0.017}, {"water": "Sudan", "X1992": 0.017, "X2002": 0.017}, {"water": "Sweden", "X1992": 0.017, "X2002": 0.017}, {"water": "Switzerland", "X1992": 0.017, "X2002": 0.017}, {"water": "Taiwan", "X1992": 0.017, "X2002": 0.017}, {"water": "Tajikistan", "X1992": 0.017, "X2002": 0.017}, {"water": "Tanzania", "X1992": 0.017, "X2002": 0.017}, {"water": "Thailand", "X1992": 0.017, "X2002": 0.017}, {"water": "Togo", "X1992": 0.017, "X2002": 0.017}, {"water": "Tonga", "X1992": 0.017, "X2002": 0.017}, {"water": "Trinidad and Tobago", "X1992": 0.017, "X2002": 0.017}, {"water": "Tunisia", "X1992": 0.017, "X2002": 0.017}, {"water": "Turkey", "X1992": 0.017, "X2002": 0.017}, {"water": "Turkmenistan", "X1992": 0.017, "X2002": 0.017}, {"water": "Uganda", "X1992": 0.017, "X2002": 0.017}, {"water": "Ukraine", "X1992": 0.017, "X2002": 0.017}, {"water": "United Arab Emirates", "X1992": 0.017, "X2002": 0.017}, {"water": "United Kingdom", "X1992": 0.017, "X2002": 0.017}, {"water": "United States", "X1992": 0.017, "X2002": 0.017}, {"water": "Uruguay", "X1992": 0.017, "X2002": 0.017}, {"water": "Uzbekistan", "X1992": 0.017, "X2002": 0.017}, {"water": "Vanuatu", "X1992": 0.017, "X2002": 0.017}, {"water": "Venezuela", "X1992": 0.017, "X2002": 0.017}, {"water": "Vietnam", "X1992": 0.017, "X2002": 0.017}, {"water": "Yemen", "X1992": 0.017, "X2002": 0.017}, {"water": "Zambia", "X1992": 0.017, "X2002": 0.017}, {"water": "Zimbabwe", "X1992": 0.017, "X2002": 0.017}]]
```

Minify and prettify JSON

```
pretty_json <- toJSON(mtcars, pretty = TRUE)
```

```
# Print pretty_json
#pretty_json - style the json one above the other
```

```
# Minify pretty_json: mini_json
mini_json <- minify(pretty_json)
```

```
# Print mini_json
mini_json
```

```
## [{"mpg": 21, "cyl": 6, "disp": 160, "hp": 110, "drat": 3.9, "wt": 2.62, "qsec": 16.46, "vs": 0, "am": 1, "gear": 4, "carb": 4}]]
```

## Importing data from other sources (SAS, STATA, SPSS)

With haven package

```
library("haven") 1. read_sas()
```

2. Import from STATA

```
library("haven")
```

```
sugar <- read_dta("http://assets.datacamp.com/production/course_1478/datasets/trade.dta")
```

```
# Structure of sugar
str(sugar)
```

```
## tibble [10 x 5] (S3: tbl_df/tbl/data.frame)
## $ Date      : dbl+lbl [1:10] 10,  9,  8,  7,  6,  5,  4,  3,  2,  1
##   ..@ label      : chr "Date"
##   ..@ format.stata: chr "%9.0g"
##   ..@ labels      : Named num [1:10] 1 2 3 4 5 6 7 8 9 10
##   .. ..- attr(*, "names")= chr [1:10] "2004-12-31" "2005-12-31" "2006-12-31" "2007-12-31" ...
## $ Import     : num [1:10] 37664782 16316512 11082246 35677943 9879878 ...
##   ..- attr(*, "label")= chr "Import"
##   ..- attr(*, "format.stata")= chr "%9.0g"
## $ Weight_I   : num [1:10] 54029106 21584365 14526089 55034932 14806865 ...
##   ..- attr(*, "label")= chr "Weight_I"
##   ..- attr(*, "format.stata")= chr "%9.0g"
## $ Export     : num [1:10] 5.45e+07 1.03e+08 3.79e+07 4.85e+07 7.15e+07 ...
##   ..- attr(*, "label")= chr "Export"
##   ..- attr(*, "format.stata")= chr "%9.0g"
## $ Weight_E   : num [1:10] 9.34e+07 1.58e+08 8.80e+07 1.12e+08 1.32e+08 ...
##   ..- attr(*, "label")= chr "Weight_E"
##   ..- attr(*, "format.stata")= chr "%9.0g"
## - attr(*, "label")= chr "Written by R."
```

```

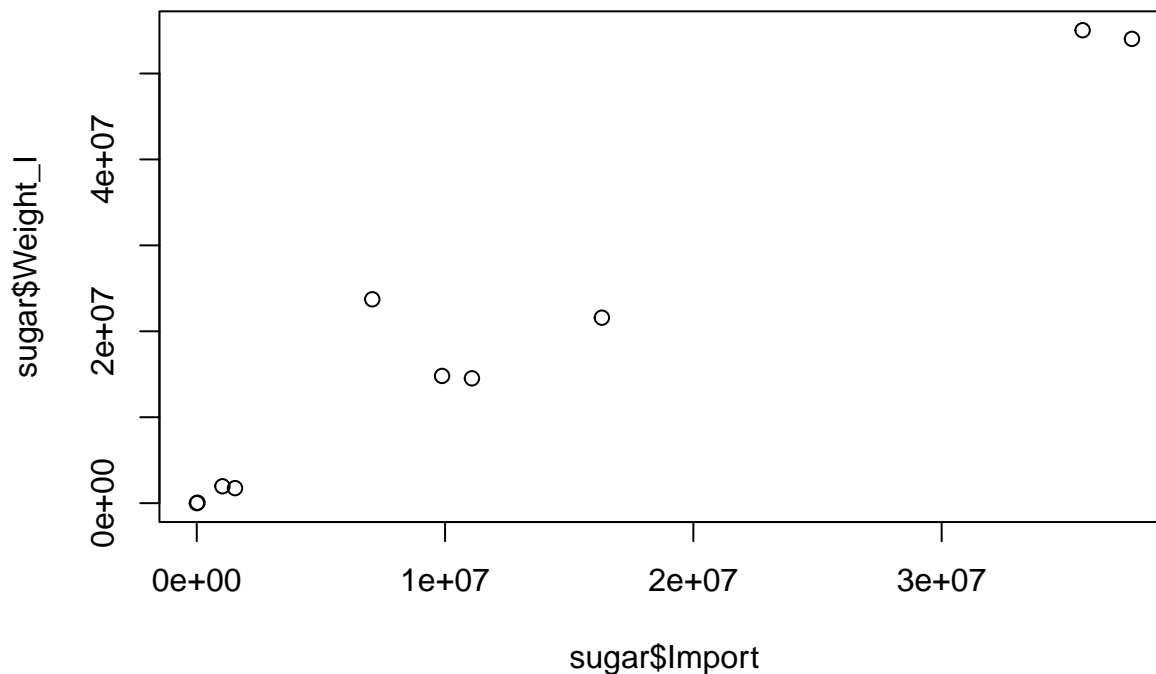
# Convert values in Date column to dates
sugar$Date <- as.Date(as_factor(sugar$Date))

# Structure of sugar again
str(sugar)

## tibble [10 x 5] (S3: tbl_df/tbl/data.frame)
##  $ Date      : Date[1:10], format: "2013-12-31" "2012-12-31" ...
##  $ Import    : num [1:10] 37664782 16316512 11082246 35677943 9879878 ...
##    .. attr(*, "label")= chr "Import"
##    .. attr(*, "format.stata")= chr "%9.0g"
##  $ Weight_I: num [1:10] 54029106 21584365 14526089 55034932 14806865 ...
##    .. attr(*, "label")= chr "Weight_I"
##    .. attr(*, "format.stata")= chr "%9.0g"
##  $ Export    : num [1:10] 5.45e+07 1.03e+08 3.79e+07 4.85e+07 7.15e+07 ...
##    .. attr(*, "label")= chr "Export"
##    .. attr(*, "format.stata")= chr "%9.0g"
##  $ Weight_E: num [1:10] 9.34e+07 1.58e+08 8.80e+07 1.12e+08 1.32e+08 ...
##    .. attr(*, "label")= chr "Weight_E"
##    .. attr(*, "format.stata")= chr "%9.0g"
##  - attr(*, "label")= chr "Written by R."

plot(sugar$Import, sugar$Weight_I)

```



3. Import data from SPSS with `read_sav()`



```

# Import SPSS data from the URL: work
work <- read_sav("http://s3.amazonaws.com/assets.datacamp.com/production/course_1478/datasets/employee.")

summary(work$GENDER)

##      Length      Class      Mode 
##      474 character character 

# Convert work$GENDER to a factor
work$GENDER <- as_factor(work$GENDER)
summary(work$GENDER)

## Female      Male 
##      216      258 

```

### With foreign package

1. From STATA read.dta("florida.dta")

The arguments you will use most often are **convert.dates**, **convert.factors**, **missing.type** and **convert.underscore**

2. From SPSS read.spss("international.sav", to.data.frame=TRUE) boxplot(demo\$gdp)