Data Wrangling: February 7, 2018

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In today’s class we will cover the various data (or “object”) structures in R. We will cover the following objects:

* scalar (which are really vectors of length 1)
* vector
* matrices (and “arrays”)
* data frame
* list
* factors

**NOTE:** To learn more about data types in R see:

* Quick-R Pages on datatypes <http://www.statmethods.net/input/datatypes.html>
* Jenny Bryan’s slides on R Objects <https://speakerdeck.com/jennybc/simple-view-of-r-objects>

We will also cover the following useful functions for checking/reviewing the data types of objects in R:

* summary()
  + summary is a generic function used to produce result summaries of the results of various model fitting functions. The function invokes particular methods which depend on the class of the first argument.
* str()
  + Compactly display the internal structure of an R object, a diagnostic function and an alternative to summary (and to some extent, dput). Ideally, only one line for each ‘basic’ structure is displayed. It is especially well suited to compactly display the (abbreviated) contents of (possibly nested) lists. The idea is to give reasonable output for any R object. It calls args for (non-primitive) function objects.
* class()
  + R possesses a simple generic function mechanism which can be used for an object-oriented style of programming. Method dispatch takes place based on the class of the first argument to the generic function.
* mode()
  + Get or set the type or storage mode of an object.
* typeof()
  + typeof determines the (R internal) type or storage mode of any object
* attributes()
  + These functions access an object’s attributes. The first form below returns the object’s attribute list. The replacement forms uses the list on the right-hand side of the assignment as the object’s attributes (if appropriate).

**NOTE:** Many times if you get an error trying to run a function in R it will be because the object you have put into the function is of the wrong class. You’ll get something warning you that there is a problem or that you have a class type mismatch.

**EXAMPLE**: Read the help pages for the lm() function which is used to fit a linear model. the usage is

lm(formula, data, subset, weights, na.action,  
 method = "qr", model = TRUE, x = FALSE, y = FALSE, qr = TRUE,  
 singular.ok = TRUE, contrasts = NULL, offset, ...)

where:

* the 1st argument formula is assumed to an object of class “formula” - which we’ll discuss later this semester
* the second argument data is assumed to a data frame or list
* the 3rd argument subset is assumed to be a vector … and so on … ALWAYS read the help pages for the function and see what types of objects it wants

## Let’s create some data/objects

### Create 5 scalar values.

a <- 3  
b <- 1  
c <- 5  
d <- 4.5  
e <- 6.234

### Create some objects

We could use these individual values to create a single vector containing these 5 values. To do this we’ll use the c() “combine values” function. We can either do it using the object names "a" "b" "c" "d" "e" or we can simply use the values themselves.

v1 <- c(a,b,c,d,e)  
v2 <- c(3,1,5,4.5,6.234)

Let’s make a couple more vectors with 5 elements. Let’s make one with characters/letters and another one with TRUE/FALSE (or T/F) logical values and a third with only whole numbers.

v.char <- c("blue","green","red","yellow","blue")  
v.log <- c(T,F,F,T,F)  
v.int <- c(2002,2004,2006,2008,2010)

### Look at the properties of the objects

Look at the class of each object

class(v1)

## [1] "numeric"

class(v2)

## [1] "numeric"

class(v.char)

## [1] "character"

class(v.log)

## [1] "logical"

class(v.int)

## [1] "numeric"

Look at the mode

mode(v1)

## [1] "numeric"

mode(v2)

## [1] "numeric"

mode(v.char)

## [1] "character"

mode(v.log)

## [1] "logical"

mode(v.int)

## [1] "numeric"

Look at the typeof each object

typeof(v1)

## [1] "double"

typeof(v2)

## [1] "double"

typeof(v.char)

## [1] "character"

typeof(v.log)

## [1] "logical"

typeof(v.int)

## [1] "double"

**NOTE:** Notice that v.int is a numeric vector not an integer. We can force this to a true integer vector using is.integer() to test to see if R thinks is it is an integer vector. If not, then we can use as.integer() to force it.

is.integer(v.int)

## [1] FALSE

v.int2 <- as.integer(v.int)  
is.integer(v.int2)

## [1] TRUE

### Object types and changes

Here are some useful functions for checking your object types and for possibly changing them as needed.

* is.character()
* is.numeric()
* is.integer()
* is.data.frame()
* is.double()
* is.list()
* is.factor()
* and there are more - just look in help for the various is.xxx() functions.

Most of these have associated as.xx() functions to help you move back and forth between object types ~ *sometimes*.

* as.character()
* as.numeric()
* as.integer()
* as.data.frame()
* as.double()
* as.list()
* as.factor()
* and there are more - just look in help for the various as.xxx() functions.

### Make a data.frame

Thus far, all of our vectors are of the same length, but they are of different type. The best data object to hold (a) vectors of the same length and (b) of different data types is a data.frame. The data.frame object is the best form of **TIDY** data where each ROW is a single CASE (or SUBJECT) and each COLUMN is a feature, measurement, or piece of information about that case (i.e. the COLUMNS are the VARIABLES).

Let’s combine v2, v.char, v.int2, and v.log into a data.frame. The easiest way to do this is using the data.frame() function. Then let’s run

* str() to see the structure listing
* summary() to see what summary stats we get for each column since they are different data types
* let’s also look at the class(),
* mode(),
* typeof() and
* attributes for the newly created df data.frame object

# create a data.frame object called "df"  
# combining 4 vectors v2, v.char, v.int2, v.log  
df <- data.frame(v2,v.char,v.int2,v.log)  
  
# look at the structure  
str(df)

## 'data.frame': 5 obs. of 4 variables:  
## $ v2 : num 3 1 5 4.5 6.23  
## $ v.char: Factor w/ 4 levels "blue","green",..: 1 2 3 4 1  
## $ v.int2: int 2002 2004 2006 2008 2010  
## $ v.log : logi TRUE FALSE FALSE TRUE FALSE

# look at the other properties of "df"  
class(df)

## [1] "data.frame"

mode(df)

## [1] "list"

typeof(df)

## [1] "list"

attributes(df)

## $names  
## [1] "v2" "v.char" "v.int2" "v.log"   
##   
## $row.names  
## [1] 1 2 3 4 5  
##   
## $class  
## [1] "data.frame"

### Attributes of a data.frame

Notice that when we ran attributes(df) there were 3 pieces of information provided:

* $names
* $row.names
* and $class

The names attributes is really helpful for labeling and selecting specific COLUMNS or VARIABLES in our new dataset “df”. There is a function names() that is useful for (a) finding out what the column/variables names are and for (b) changing the variable names if we need to.

Right now the column names are not helpful. The names simply reflect the previous vector names.

names(df)

## [1] "v2" "v.char" "v.int2" "v.log"

Let’s change the names to something more interesting.

* for “v2” we’ll change that to “avgvisit” (hypothetical average number of visits to somewhere)
* for “v.char” change that to “color” (hypothetical color categories for plotting later)
* for “v.int2” change to “year” (hypothetical year the data was collected)
* and for “v.log” change to “valid” (hypothetical indicator for whether the data is validated or not)

To do this we’ll use the c() combine function and assemble these new labels to rename the current column names. Run the names(df) before applying the new names, then assign the new names, and run names(df) again to see/check that the column names have been updated.

# see the original variable/column names  
names(df)

## [1] "v2" "v.char" "v.int2" "v.log"

# assign the new variable/column names  
names(df) <- c("avgvisit","color","year","valid")  
  
# see that the variable/column names have updated  
names(df)

## [1] "avgvisit" "color" "year" "valid"

### View/Print the dataset (as a table)

Since this is such a small dataset with 5 ROWS and 4 COLUMNS we can easily “view” it by printing it in a table. For this we’ll use the kable() function from the knitr package. To call a specific function in a specific package, you list the package first followed by 2 colons :: and then the function.

**NOTE:** The kable() function makes pretty good tables in HTML, DOCX and PDF formats for objects that are data.frame or a matrix. We’ll learn more about the kable() function throughout the semester… In the example below, I also added a caption for the table.

knitr::kable(df,  
 caption="View the 'df' object")

View the ‘df’ object

|  |  |  |  |
| --- | --- | --- | --- |
| avgvisit | color | year | valid |
| 3.000 | blue | 2002 | TRUE |
| 1.000 | green | 2004 | FALSE |
| 5.000 | red | 2006 | FALSE |
| 4.500 | yellow | 2008 | TRUE |
| 6.234 | blue | 2010 | FALSE |

## Worked Example from the UCI Data Repository

The following dataset comes from the [UCI Data Repository](http://archive.ics.uci.edu/ml/). The dataset we’ll use is the Contraceptive Method Choice dataset. The information on this dataset is provided at <http://archive.ics.uci.edu/ml/datasets/Contraceptive+Method+Choice>. If you click on the “Data Folder” you can download the RAW data cmc.data which is a comma delimited format dataset (i.e. it is a CSV formatted file) and the description of the data included, the variable names and associated codes for the values included which is in the cmc.names file. See “Data Folder”" at <http://archive.ics.uci.edu/ml/machine-learning-databases/cmc/>

### Read-in data

**NOTE:** Download the 2 files from the UCI Data Repository for the Contraceptive Method Choice and put them in the directory where you have this RMD rmarkdown file.

# load the tidyverse package(s)  
library(tidyverse)

## Loading tidyverse: ggplot2  
## Loading tidyverse: tibble  
## Loading tidyverse: tidyr  
## Loading tidyverse: readr  
## Loading tidyverse: purrr  
## Loading tidyverse: dplyr

## Conflicts with tidy packages ----------------------------------------------

## filter(): dplyr, stats  
## lag(): dplyr, stats

# read in the comma delimited (CSV) formatted dataset  
# \*\*NOTE\*\*: This dataset does NOT have the column  
# names as the 1st row of the file. We will assign the   
# column names below.  
cmc <- read\_csv("cmc.data", col\_names=FALSE)

## Parsed with column specification:  
## cols(  
## X1 = col\_integer(),  
## X2 = col\_integer(),  
## X3 = col\_integer(),  
## X4 = col\_integer(),  
## X5 = col\_integer(),  
## X6 = col\_integer(),  
## X7 = col\_integer(),  
## X8 = col\_integer(),  
## X9 = col\_integer(),  
## X10 = col\_integer()  
## )

### Apply the codebook - variable names and coding used

Apply variable names to the 10 columns of data in cmc.

# the variable names before we change them  
# Notice that the columns names are very generic  
# X1, X2, ..., X10  
names(cmc)

## [1] "X1" "X2" "X3" "X4" "X5" "X6" "X7" "X8" "X9" "X10"

# assign new variables names to the 10 columns  
names(cmc) <- c("WifeAge", "WifeEd", "HusbEd", "NumChild",  
 "WifeRel", "WifeWork", "HusbOcc", "SOLindex",   
 "Media", "Contraceptive")  
  
# see the updated column/variable names  
names(cmc)

## [1] "WifeAge" "WifeEd" "HusbEd" "NumChild"   
## [5] "WifeRel" "WifeWork" "HusbOcc" "SOLindex"   
## [9] "Media" "Contraceptive"

The next code chunk is to add the labels for “factor” levels for some of the variables (i.e. we are creating factors).

**WARNING**: Notice I’m overwriting the variables and changing them from integers to factors which have different properties as you’ll see below. If you want to keep the original integer variables, you could simply give the new facotr variable a new name. For example you could write

cmc$WifeEd.f <- factor(cmc$WifeEd,  
 levels = c(1,2,3,4),  
 labels = c("low","med low","med

and this would append a new column onto the cmc dataset that is the “factor” type version of Wife’s Education. For now, use the code below to update all of the variables.

# notice that Wife Education is currently of "Integer" class type  
# see what you get from the summary() function  
class(cmc$WifeEd)

## [1] "integer"

summary(cmc$WifeEd)

## Min. 1st Qu. Median Mean 3rd Qu. Max.   
## 1.000 2.000 3.000 2.959 4.000 4.000

# update Wife Education as a factor, assign the   
# levels and the labels for each level  
cmc$WifeEd <- factor(cmc$WifeEd,  
 levels = c(1,2,3,4),  
 labels = c("low","med low","med high","high"))  
  
# repeat the above commands to see what changes  
class(cmc$WifeEd)

## [1] "factor"

summary(cmc$WifeEd)

## low med low med high high   
## 152 334 410 577

# do the remaining variables  
cmc$HusbEd <- factor(cmc$HusbEd,  
 levels = c(1,2,3,4),  
 labels = c("low","med low","med high","high"))  
  
cmc$WifeRel <- factor(cmc$WifeRel,  
 levels = c(0,1),  
 labels = c("Non-Islam","Islam"))  
  
# Note: The documentation does state that  
# 0=yes and 1=no which seems incorrect...  
cmc$WifeWork <- factor(cmc$WifeWork,  
 levels = c(0,1),  
 labels = c("Yes","No"))  
  
cmc$HusbOcc <- factor(cmc$HusbOcc,  
 levels = c(1,2,3,4),  
 labels = c("1","2","3","4"))  
  
cmc$SOLindex <- factor(cmc$SOLindex,  
 levels = c(1,2,3,4),  
 labels = c("low","med low","med high","high"))  
  
cmc$Media <- factor(cmc$Media,  
 levels = c(0,1),  
 labels = c("Good","Not Good"))  
  
cmc$Contraceptive <- factor(cmc$Contraceptive,  
 levels = c(1,2,3),  
 labels = c("No-use","Long-term","Short-term"))

### Look at a subset of the data

head(cmc)

## # A tibble: 6 x 10  
## WifeAge WifeEd HusbEd NumChild WifeRel WifeWork HusbOcc SOLindex  
## <int> <fctr> <fctr> <int> <fctr> <fctr> <fctr> <fctr>  
## 1 24 med low med high 3 Islam No 2 med high  
## 2 45 low med high 10 Islam No 3 high  
## 3 43 med low med high 7 Islam No 3 high  
## 4 42 med high med low 9 Islam No 3 med high  
## 5 36 med high med high 8 Islam No 3 med low  
## 6 19 high high 0 Islam No 3 med high  
## # ... with 2 more variables: Media <fctr>, Contraceptive <fctr>

### Print this subset using knitr::kable()

knitr::kable(head(cmc))

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| WifeAge | WifeEd | HusbEd | NumChild | WifeRel | WifeWork | HusbOcc | SOLindex | Media | Contraceptive |
| 24 | med low | med high | 3 | Islam | No | 2 | med high | Good | No-use |
| 45 | low | med high | 10 | Islam | No | 3 | high | Good | No-use |
| 43 | med low | med high | 7 | Islam | No | 3 | high | Good | No-use |
| 42 | med high | med low | 9 | Islam | No | 3 | med high | Good | No-use |
| 36 | med high | med high | 8 | Islam | No | 3 | med low | Good | No-use |
| 19 | high | high | 0 | Islam | No | 3 | med high | Good | No-use |

### Summarize the dataset

**NOTICE** that Wife’s Age and Number of Children are now the only “numeric” “integer” variables - these are the only ones for which we get summary statistics. All the remaining variables are “factors” so we only get the frequencies for each category.

summary(cmc)

## WifeAge WifeEd HusbEd NumChild   
## Min. :16.00 low :152 low : 44 Min. : 0.000   
## 1st Qu.:26.00 med low :334 med low :178 1st Qu.: 1.000   
## Median :32.00 med high:410 med high:352 Median : 3.000   
## Mean :32.54 high :577 high :899 Mean : 3.261   
## 3rd Qu.:39.00 3rd Qu.: 4.000   
## Max. :49.00 Max. :16.000   
## WifeRel WifeWork HusbOcc SOLindex Media   
## Non-Islam: 220 Yes: 369 1:436 low :129 Good :1364   
## Islam :1253 No :1104 2:425 med low :229 Not Good: 109   
## 3:585 med high:431   
## 4: 27 high :684   
##   
##   
## Contraceptive  
## No-use :629   
## Long-term :333   
## Short-term:511   
##   
##   
##

### Computing stats on factors

Suppose you wanted to know the mean education level of the Huband’s in this dataset. We can use the as.numeric() function to convert the variable and then run a mean() on it. We’ll do more on facotrs later this semester.

mean(as.numeric(cmc$HusbEd))

## [1] 3.429735

### Select various elements

# select the first 10 rows  
cmc\_first10 <- cmc[1:10,]  
cmc\_first10

## # A tibble: 10 x 10  
## WifeAge WifeEd HusbEd NumChild WifeRel WifeWork HusbOcc SOLindex  
## <int> <fctr> <fctr> <int> <fctr> <fctr> <fctr> <fctr>  
## 1 24 med low med high 3 Islam No 2 med high  
## 2 45 low med high 10 Islam No 3 high  
## 3 43 med low med high 7 Islam No 3 high  
## 4 42 med high med low 9 Islam No 3 med high  
## 5 36 med high med high 8 Islam No 3 med low  
## 6 19 high high 0 Islam No 3 med high  
## 7 38 med low med high 6 Islam No 3 med low  
## 8 21 med high med high 1 Islam Yes 3 med low  
## 9 27 med low med high 3 Islam No 3 high  
## 10 45 low low 8 Islam No 2 med low  
## # ... with 2 more variables: Media <fctr>, Contraceptive <fctr>

# select the first 2 columns  
cmc\_cols12 <- cmc[,1:2]  
cmc\_cols12

## # A tibble: 1,473 x 2  
## WifeAge WifeEd  
## <int> <fctr>  
## 1 24 med low  
## 2 45 low  
## 3 43 med low  
## 4 42 med high  
## 5 36 med high  
## 6 19 high  
## 7 38 med low  
## 8 21 med high  
## 9 27 med low  
## 10 45 low  
## # ... with 1,463 more rows

# select the value in 5th row, 6th column  
cmc\_row5col6 <- cmc[5,6]  
cmc\_row5col6

## # A tibble: 1 x 1  
## WifeWork  
## <fctr>  
## 1 No

### Select Rows

Use the following code to find the number of subjects (number of rows) where the wife’s age is > 40.

# find the dimensions of the cmc dataset  
dim(cmc)

## [1] 1473 10

# the 1st element is the number of rows, the 2nd   
# is the number of columns - just look at number of rows  
dim(cmc)[1]

## [1] 1473

# select only the rows where WifeAge > 40  
cmc\_wife\_gt40 <- cmc[cmc$WifeAge>40,]  
  
# get number of rows of this subset  
dim(cmc\_wife\_gt40)[1]

## [1] 306

### Select Columns

We can select columns by position or by name. Select the columns for the Wife variables, which are in columns 1, 2, 5, 6 - these variables are:

* WifeAge
* WifeEd
* WifeRel
* WifeWork

# select these columns by position  
cmc\_wifeOnly <- cmc[,c(1,2,5,6)]  
head(cmc\_wifeOnly)

## # A tibble: 6 x 4  
## WifeAge WifeEd WifeRel WifeWork  
## <int> <fctr> <fctr> <fctr>  
## 1 24 med low Islam No  
## 2 45 low Islam No  
## 3 43 med low Islam No  
## 4 42 med high Islam No  
## 5 36 med high Islam No  
## 6 19 high Islam No

# select these columns by variable names  
keepvars <- c("WifeAge", "WifeEd", "WifeRel", "WifeWork")  
cmc\_wifeOnly <- cmc[,keepvars]  
head(cmc\_wifeOnly)

## # A tibble: 6 x 4  
## WifeAge WifeEd WifeRel WifeWork  
## <int> <fctr> <fctr> <fctr>  
## 1 24 med low Islam No  
## 2 45 low Islam No  
## 3 43 med low Islam No  
## 4 42 med high Islam No  
## 5 36 med high Islam No  
## 6 19 high Islam No

## TIDY Data

We’ll use the following functions as we go through this semester, for now, let’s review the following package(s) and the whole TIDYVERSE which is very helpful for working with data in many different formats and structures.

* The [“TIDYVERSE”](http://tidyverse.org/)
* [“Tidy Data” paper by Hadley Wickham; Journal of Statistical Software; v.59 (2014)](https://www.jstatsoft.org/article/view/v059i10)
  + **NOTE**: The original code and datasets can be reviewed at the Github repository for this paper at <https://github.com/hadley/tidy-data>.
* [“R for Data Science” book by Hadley Wickham; O’Reilly Media Inc. (2017) - Part II on Data Wrangling](http://r4ds.had.co.nz/wrangle-intro.html)