

Data Classes

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Data Classes:

- ▶ One dimensional classes ('vectors'):
 - ▶ Character: strings or individual characters, quoted
 - ▶ Numeric: any real number(s)
 - ▶ Integer: any integer(s)/whole numbers
 - ▶ Factor: categorical/qualitative variables
 - ▶ Logical: variables composed of TRUE or FALSE
 - ▶ Date/POSIXct: represents calendar dates and times

Character and numeric

We have already covered character and numeric classes.

```
class(c("Andrew", "Jaffe"))
```

```
## [1] "character"
```

```
class(c(1, 4, 7))
```

```
## [1] "numeric"
```

Integer

Integer is a special subset of numeric that contains only whole numbers

A sequence of numbers is an example of the integer class

```
x = seq(from = 1, to = 5) # seq() is a function  
x
```

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

```
class(x)
```

```
## [1] "integer"
```

Integer

The colon `:` is a shortcut for making sequences of numbers

It makes consecutive integer sequence from `[num1]` to `[num2]` by 1

```
1:5
```

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

Logical

`logical` is a class that only has two possible elements: `TRUE` and `FALSE`

```
x = c(TRUE, FALSE, TRUE, TRUE, FALSE)
class(x)
```

```
## [1] "logical"
```

`sum()` and `mean()` work on `logical` vectors - they return the total and proportion of `TRUE` elements, respectively.

Logical

Note that logical elements are NOT in quotes.

```
z = c(TRUE, FALSE, TRUE, FALSE)
class(z)
```

```
## [1] "character"
```

Factor

factor are special character vectors where the elements have pre-defined groups or 'levels'. You can think of these as qualitative or categorical variables:

```
x = factor(c("boy", "girl", "girl", "boy", "girl"))  
x
```

```
## [1] boy  girl girl boy  girl  
## Levels: boy girl
```

```
class(x)
```

```
## [1] "factor"
```

Note that levels are, by default, alphabetical or alphanumerical order.

Factors

Factors are used to represent categorical data, and can also be used for ordinal data (ie categories have an intrinsic ordering)

Note that R reads in character strings as factors by default in functions like `read.table()`

'The function `factor` is used to encode a vector as a factor (the terms 'category' and 'enumerated type' are also used for factors). If argument `ordered` is `TRUE`, the factor levels are assumed to be ordered.'

```
factor(x = character(), levels, labels = levels,  
       exclude = NA, ordered = is.ordered(x))
```

Factors

Suppose we have a vector of case-control status

```
cc = factor(c("case", "case", "case",  
              "control", "control", "control"))  
cc
```

```
## [1] case    case    case    control control control  
## Levels: case control
```

```
levels(cc) = c("control", "case")  
cc
```

```
## [1] control control control case    case    case  
## Levels: control case
```

Factors

Note that the levels are alphabetically ordered by default. We can also specify the levels within the factor call

```
factor(c("case","case","case","control",  
        "control","control"),  
       levels =c("control","case") )
```

```
## [1] case    case    case    control control control  
## Levels: control case
```

```
factor(c("case","case","case","control",  
        "control","control"),  
       levels =c("control","case"), ordered=TRUE)
```

```
## [1] case    case    case    control control control  
## Levels: control < case
```

Factors

Factors can be converted to numeric or character very easily

```
x = factor(c("case","case","case","control",  
            "control","control"),  
           levels =c("control","case") )  
as.character(x)
```

```
## [1] "case"      "case"      "case"      "control" "control" "control"
```

```
as.numeric(x)
```

```
## [1] 2 2 2 1 1 1
```

Factors

However, you need to be careful modifying the labels of existing factors, as its quite easy to alter the meaning of the underlying data.

```
xCopy = x  
levels(xCopy) = c("case", "control") # wrong way  
xCopy
```

```
## [1] control control control case      case      case  
## Levels: case control
```

```
as.character(xCopy) # labels switched
```

```
## [1] "control" "control" "control" "case"      "case"      "c
```

```
as.numeric(xCopy)
```

```
## [1] 2 2 2 1 1 1
```

Creating categorical variables

the `rep()` ["repeat"] function is useful for creating new variables

```
bg = rep(c("boy", "girl"), each=50)
head(bg)
```

```
## [1] "boy" "boy" "boy" "boy" "boy" "boy"
```

```
bg2 = rep(c("boy", "girl"), times=50)
head(bg2)
```

```
## [1] "boy" "girl" "boy" "girl" "boy" "girl"
```

```
length(bg)==length(bg2)
```

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

Creating categorical variables

One frequently-used tool is creating categorical variables out of continuous variables, like generating quantiles of a specific continuously measured variable.

A general function for creating new variables based on existing variables is the `ifelse()` function, which “returns a value with the same shape as test which is filled with elements selected from either yes or no depending on whether the element of test is TRUE or FALSE.”

```
ifelse(test, yes, no)
```

```
# test: an object which can be coerced  
      to logical mode.
```

```
# yes: return values for true elements of test.
```

```
# no: return values for false elements of test.
```

Charm City Circulator data

Please download the Charm City Circulator data:

http://www.aejaffe.com/winterR_2016/data/Charm_City_Circulator_Ridership.csv

```
circ = read.csv("http://www.aejaffe.com/winterR_2016/data/C  
              header=TRUE,as.is=TRUE)
```


Creating categorical variables

For example, we can create a new variable that records whether daily ridership on the Circulator was above 10,000.

```
hi_rider = ifelse(circ$daily > 10000, "high", "low")
hi_rider = factor(hi_rider, levels = c("low", "high"))
head(hi_rider)
```

```
## [1] low low low low low low
## Levels: low high
```

```
table(hi_rider)
```

```
## hi_rider
##  low high
##  740  282
```

Creating categorical variables

You can also nest `ifelse()` within itself to create 3 levels of a variable.

```
riderLevels = ifelse(circ$daily < 10000, "low",  
                    ifelse(circ$daily > 20000,  
                          "high", "med"))  
riderLevels = factor(riderLevels,  
                    levels = c("low","med","high"))  
head(riderLevels)
```

```
## [1] low low low low low low  
## Levels: low med high
```

```
table(riderLevels)
```

```
## riderLevels  
##   low   med  high  
##  740  280    2
```

Creating categorical variables

However, it's much easier to use `cut()` to create categorical variables from continuous variables.

'cut divides the range of `x` into intervals and codes the values in `x` according to which interval they fall. The leftmost interval corresponds to level one, the next leftmost to level two and so on.'

```
cut(x, breaks, labels = NULL, include.lowest = FALSE,  
    right = TRUE, dig.lab = 3,  
    ordered_result = FALSE, ...)
```

Creating categorical variables

`x`: a numeric vector which is to be converted to a factor by cutting.

`breaks`: either a numeric vector of two or more unique cut points or a single number (greater than or equal to 2) giving the number of intervals into which `x` is to be cut.

`labels`: labels for the levels of the resulting category. By default, labels are constructed using “(a,b]” interval notation. If `labels = FALSE`, simple integer codes are returned instead of a factor.

Cut

Now that we know more about factors, `cut()` will make more sense:

```
x = 1:100  
cx = cut(x, breaks=c(0,10,25,50,100))  
head(cx)
```

```
## [1] (0,10] (0,10] (0,10] (0,10] (0,10] (0,10]  
## Levels: (0,10] (10,25] (25,50] (50,100]
```

```
table(cx)
```

```
## cx  
##   (0,10]  (10,25]  (25,50]  (50,100]  
##        10        15        25        50
```

We can also leave off the labels

```
cx = cut(x, breaks=c(0,10,25,50,100), labels=FALSE)  
head(cx)
```

Date

You can convert date-like strings in the Date class
(<http://www.statmethods.net/input/dates.html> for more info)

```
head(sort(circ$date))
```

```
## [1] "01/01/2011" "01/01/2012" "01/01/2013" "01/02/2011"  
## [6] "01/02/2013"
```

```
circ$newDate <- as.Date(circ$date, "%m/%d/%Y") # creating  
head(circ$newDate)
```

```
## [1] "2010-01-11" "2010-01-12" "2010-01-13" "2010-01-14"  
## [6] "2010-01-16"
```

```
range(circ$newDate)
```

```
## [1] "2010-01-11" "2013-03-01"
```

Date

However, the lubridate package is much easier for generating explicit dates:

```
library(lubridate) # great for dates!  
suppressPackageStartupMessages(library(dplyr))  
circ = mutate(circ, newDate2 = mdy(date))  
head(circ$newDate2)
```

```
## [1] "2010-01-11 UTC" "2010-01-12 UTC" "2010-01-13 UTC" "  
## [5] "2010-01-15 UTC" "2010-01-16 UTC"
```

```
range(circ$newDate2)
```

```
## [1] "2010-01-11 UTC" "2013-03-01 UTC"
```

POSIXct

The POSIXct class can encode time information

```
theTime = Sys.time()  
theTime
```

```
## [1] "2016-01-06 19:54:29 EST"
```

```
class(theTime)
```

```
## [1] "POSIXct" "POSIXt"
```

```
theTime + 5000
```

```
## [1] "2016-01-06 21:17:49 EST"
```

Note it's like a more general date format.

Data Classes:

- ▶ Two dimensional classes:
 - ▶ `data.frame`: traditional 'Excel' spreadsheets
 - ▶ Each column can have a different class, from above
 - ▶ Matrix: two-dimensional data, composed of rows and columns. Unlike data frames, the entire matrix is composed of one R class, e.g. all numeric or all characters.

Matrices

```
n = 1:9  
n
```

```
## [1] 1 2 3 4 5 6 7 8 9
```

```
mat = matrix(n, nrow = 3)  
mat
```

```
##      [,1] [,2] [,3]  
## [1,]    1    4    7  
## [2,]    2    5    8  
## [3,]    3    6    9
```

Matrix (and Data frame) Functions

These are in addition to the previous useful vector functions:

- ▶ `nrow()` displays the number of rows of a matrix or data frame
- ▶ `ncol()` displays the number of columns
- ▶ `dim()` displays a vector of length 2: # rows, # columns
- ▶ `colnames()` displays the column names (if any) and
`rownames()` displays the row names (if any)

Data Selection

Matrices have two “slots” you can use to select data, which represent rows and columns, that are separated by a comma, so the syntax is `matrix[row,column]`. Note you cannot use `dplyr` functions on matrices.

```
mat[1, 1] # individual entry: row 1, column 1
```

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

```
mat[1, ] # first row
```

```
## [1] 1 4 7
```

```
mat[, 1] # first columns
```

```
## [1] 1 2 3
```

Data Selection

Note that the class of the returned object is no longer a matrix

```
class(mat[1, ])
```

```
## [1] "integer"
```

```
class(mat[, 1])
```

```
## [1] "integer"
```

Data Frames

To review, the `data.frame` is the other two dimensional variable class.

Again, data frames are like matrices, but each column is a vector that can have its own class. So some columns might be `character` and others might be `numeric`, while others maybe a `factor`.

Data Frames versus Matrices

You will likely use `data.frame` class for a lot of data cleaning and analysis. However, some operations that rely on matrix multiplication (like performing many linear regressions) are (much) faster with matrices. Also, as we will touch on later, some functions for iterating over data will return the matrix class, or will be placed in empty matrices that can then be converted to `data.frames`

Data Frames versus Matrices

There is also additional summarization functions for matrices (and not data.frames) in the `matrixStats` package, like `rowMins()`, `colMaxs()`, etc.

```
library(matrixStats,quietly = TRUE)
avgs = select(circ, ends_with("Average"))
rowMins(as.matrix(avgs),na.rm=TRUE)[500:510]
```

```
## [1] 3538.5 3402.5 3862.5 3347.5 2837.5 2704.0 3138.5 32
## [11] 3046.0
```


Data Classes

Extensions of “normal” data classes:

- ▶ N-dimensional classes:
 - ▶ Arrays: any extension of matrices with more than 2 dimensions, e.g. 3x3x3 cube
 - ▶ Lists: more flexible container for R objects.

Arrays

These are just more flexible matrices - you should just be made aware of them as some functions return objects of this class, for example, cross tabulating over more than 2 variables and the `tapply` function.

Arrays

Selecting from arrays is similar to matrices, just with additional commas for the additional slots.

```
ar = array(1:27, c(3,3,3))  
ar[, , 1]
```

```
##           [,1] [,2] [,3]  
## [1,]         1     4     7  
## [2,]         2     5     8  
## [3,]         3     6     9
```

```
ar[, 1, ]
```

```
##           [,1] [,2] [,3]  
## [1,]         1    10    19  
## [2,]         2    11    20  
## [3,]         3    12    21
```

Lists

- ▶ One other data type that is the most generic are lists.
- ▶ Can be created using `list()`
- ▶ Can hold vectors, strings, matrices, models, list of other list, lists upon lists!
- ▶ Can reference data using `$` (if the elements are named), or using `[]`, or `[[]]`

```
> mylist <- list(letters=c("A", "b", "c"),  
+               numbers=1:3, matrix(1:25, ncol=5))
```

List Structure

```
> head(mylist)
```

```
$letters
```

```
[1] "A" "b" "c"
```

```
$numbers
```

```
[1] 1 2 3
```

```
[[3]]
```

	[,1]	[,2]	[,3]	[,4]	[,5]
[1,]	1	6	11	16	21
[2,]	2	7	12	17	22
[3,]	3	8	13	18	23
[4,]	4	9	14	19	24
[5,]	5	10	15	20	25

List referencing

```
> mylist[1] # returns a list
```

```
$letters  
[1] "A" "b" "c"
```

```
> mylist["letters"] # returns a list
```

```
$letters  
[1] "A" "b" "c"
```

List referencing

```
> mylist[[1]] # returns the vector 'letters'
```

```
[1] "A" "b" "c"
```

```
> mylist$letters # returns vector
```

```
[1] "A" "b" "c"
```

```
> mylist[["letters"]] # returns the vector 'letters'
```

```
[1] "A" "b" "c"
```

List referencing

You can also select multiple lists with the single brackets.

```
> mylist[1:2] # returns a list
```

```
$letters
```

```
[1] "A" "b" "c"
```

```
$numbers
```

```
[1] 1 2 3
```


List referencing

You can also select down several levels of a list at once

```
> mylist$letters[1]
```

```
[1] "A"
```

```
> mylist[[2]][1]
```

```
[1] 1
```

```
> mylist[[3]][1:2,1:2]
```

	[,1]	[,2]
[1,]	1	6
[2,]	2	7

Splitting Data Frames

The `split()` function is useful for splitting `data.frames`

“`split` divides the data in the vector `x` into the groups defined by `f`. The replacement forms replace values corresponding to such a division. `unsplit` reverses the effect of `split`.”

```
> dayList = split(circ,circ$day)
```

Splitting Data Frames

Here is a good chance to introduce `lapply`, which performs a function within each list element:

```
> # head(dayList)
> lapply(dayList, head, n=2)
```

\$Friday

	day	date	orangeBoardings	orangeAlightings	orangeAverage	purpleBoardings	purpleAlightings	purpleAverage	greenBoardings	greenAlightings	greenAverage	bannerBoardings	bannerAlightings	bannerAverage	daily	newDate	newDate2
5	Friday	01/15/2010	1645	1643	1644	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	2010-01-15	2010-01-15
12	Friday	01/22/2010	1401	1388	1394	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	2010-01-22	2010-01-22

```
> # head(dayList)
> lapply(dayList, dim)
```

```
$Friday
[1] 164 17
```

```
$Monday
[1] 164 17
```

```
$Saturday
[1] 163 17
```

```
$Sunday
[1] 163 17
```

```
$Thursday
[1] 164 17
```

```
$Tuesday
[1] 164 17
```

General Class Information

There are two useful functions associated with practically all R classes, which relate to logically checking the underlying class (`is.CLASS_()`) and coercing between classes (`as.CLASS_()`).

We saw some examples of coercion in the past, like `as.numeric()` and `as.character()` regarding the factor class and also `as.Date()` for the date class.