# 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
$\mathrm{x}=\operatorname{seq}($ from $=1$, to $=5)$ \# seq() is a function
X
\#\# [1] 12345
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] 12345

## 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.'

$$
\begin{gathered}
\text { factor }(x=\text { character }(), \text { levels, labels }=\text { levels, } \\
\text { exclude }=N A, \text { ordered }=\text { is.ordered }(x))
\end{gathered}
$$

## 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 $(c c)=c(" c o n t r o l ", " c a s e ")$
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" "c
```

as.numeric(x)
\#\# [1] 222111

## 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.
$\mathrm{xCopy}=\mathrm{x}$
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"
as.numeric(xCopy)
\#\# [1] 222111

## Creating categorical variables

 the rep() ["repeat"] function is useful for creating new variablesbg = 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
\#\# 740282

## 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, \(\operatorname{dig} .1 \mathrm{ab}=3\),
    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 22:45:41 EST"
class(theTime)
## [1] "POSIXct" "POSIXt"
theTime + 5000
## [1] "2016-01-07 00:09:01 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
```


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] 147
mat[, 1] \# first columns
\#\# [1] 123

## 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] 3538.53402 .53862 .53347 .52837 .52704 .03138 .532
\#\# [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. $3 \times 3 \times 3$ 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 |

$\operatorname{ar}[, 1$,

| \#\# | $[, 1]$ | $[, 2]$ | $[, 3]$ |
| :--- | ---: | ---: | ---: |
| \#\# [1,] | 1 | 10 | 19 |
| \#\# [2,] | 2 | 11 | 20 |
| \#\# [3,] | 3 | 12 | 21 |

- 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] 123
[[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] 123

## 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 orang |
| :--- | ---: | ---: | ---: |
| 5 | Friday $01 / 15 / 2010$ | 1645 | 1643 |
| 12 | Friday $01 / 22 / 2010$ | 1401 | 1388 |
|  | purpleBoardings purpleAlightings | purpleAverage greenBoal |  |
| 5 | NA | NA | NA |
| 12 | NA | NA | NA | greenAlightings greenAverage bannerBoardings bannerAligh

5
12
NA
NA NA
NA
NA
NA
bannerAverage daily newDate newDate2
5

$$
\text { NA } 1644.0 \text { 2010-01-15 2010-01-15 }
$$

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

\$Friday
[1] 16417

| \$Monday |  |
| :--- | :--- |
| [1] 164 | 17 |

\$Saturday
[1] 16317
\$Sunday
[1] 16317
\$Thursday
[1] 16417
\$Tuesday

## 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.

