

Programming in Data science

Factors, Missing values and Data frames

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Factors

- ▶ Factors are used to **represent categorical data**.
- ▶ One can think of a factor as an integer vector where each integer has a label.
- ▶ Factors are important in statistical modeling.
- ▶ **Using factors** with labels **is better** than using **integers** because factors are self-describing.

Examples

1. Having a variable that has values “Male” and “Female” is better than a variable that has values 1 and 2.
2. A data field such as marital status may contain only values from single, married, separated, divorced, or widowed.

Factors

- Factor objects can be created with the `factor(..)` function.

```
1 > x <- factor(c("single", "married", "married", "↵
    single"));
2 > x
3 [1] single married married single
4 Levels: married single
```

- Levels may be predefined even if not used.

```
1 > x <- factor(c("single", "married", "married", "↵
    single"),
2 levels = c("single", "married", "divorced"));
3 > x
4 [1] single married married single
5 Levels: single married divorced
```

Factors

Sample Code

```

1  > x <- factor(c("single","married","married","↵
    single"))
2  > str(x)
3  Factor w/ 2 levels "married","single": 2 1 1 2

```

- Levels are stored in a character vector and the individual elements are actually stored as indices.

Access components of a factor

- Accessing components of a factor is very similar to that of vectors.

Sample Code

```
1 > x[3]                # access 3rd element
2 > x[c(2, 4)]          # access 2nd and 4th element
3 > x[-1]               # access all but 1st element
4 > x[c(TRUE, FALSE, FALSE, TRUE)] # using logical ←  
    vector
```

Modify Factors

- Components of a factor can be modified using simple assignments.

Sample Code

```

1  > x
2  [1] single married married single
3  Levels: single married divorced
4  > x[2] <- "divorced"      # modify second element;
5  > x
6  [1] single divorced married single
7  Levels: single married divorced

```

Modify Factors

Invalid assignment

```

1 > x[3] <- "widowed"      # cannot assign values ←
    outside levels
2 Warning message:
3 In factor(*tmp*, 3, value = "widowed") :
4 invalid factor level, NA generated
5 > x
6 [1] single    divorced <NA>      single
7 Levels: single married divorced

```

- However, cannot choose values outside of predefined levels.

```

1 > k <- factor(c("Male"))
2 > k[1] <- "Female"

```


Modify Factors

- A workaround to this is to add the value to the level first.

Sample Code

```

1  >k <- factor(c("Male"))
2  > k
3  [1] Male
4  Levels: Male
5
6  > levels(k) <- c(levels(k), "Female")
7  > k
8  [1] Male
9  Levels: Male Female

```

- Often factors will be automatically created for you when you read a dataset in using a function like `read.table(..)`.

Missing Values

- ▶ Missing values are denoted by **NA** (not available) or **NaN** (not a number) for undefined mathematical operations.
 - ▶ **NA** used to represent a value that is not known, as a placeholder.
 - ▶ NA says no result was available or the result is missing. It can be used in a matrix to fill in a value of a vector.
 - ▶ **NaN** implies a result that cannot be calculated for whatever reason.

Example – Sqrt of a negative number

```

1  sqrt(-1)
2  [1] NaN
3  Warning message:
4  In sqrt(-1) : NaNs produced

```

Missing Values

- ▶ Following functions can be used to test objects.
 - ▶ `is.na(..)` is used to test objects if they are NA
 - ▶ `is.nan(..)` is used to test for NaN
 - ▶ NA values have a class also, so there are integer NA, character NA, etc

is

```

1  ## Create a vector with NAs in it
2  x <- c(1, 2, NA, 10, 3)
3  ## Return a logical vector indicating which ←
   elements are NA
4  is.na(x)
5  [1] FALSE FALSE TRUE FALSE FALSE
6  ## Return a logical vector indicating which ←
   elements are NaN
7  is.nan(x)
8  [1] FALSE FALSE FALSE FALSE FALSE

```

Missing Values

- Why `is.na(..)` can detect NA and NaN whereas, `is.nan(..)` only detect NaN?

Sample Code

```
1  ## Now create a vector with both NA and NaN values
2  x <- c(1, 2, NaN, NA, 4)
3  is.na(x)
4  [1] FALSE FALSE TRUE TRUE FALSE
5  is.nan(x)
6  [1] FALSE FALSE TRUE FALSE FALSE
```

Missing Values

► Excluding Missing Values from Analyses

Sample Code

```
1 > x <- c(1,2,NA,3)
2 > mean(x)
3 [1] NA
4 > mean(x, na.rm=TRUE)
5 [1] 2
```

Exclude Missing Values Using `complete.cases(..)`

- ▶ Excluding Missing Values from a vector using `complete.cases(..)`

Sample Code

```

1  > vec <- c(a=c(1,2,3,4,NA), b=c(3,4,5,6,NA))
2  > vec
3  a1 a2 a3 a4 a5 b1 b2 b3 b4 b5
4   1  2  3  4 NA  3  4  5  6 NA
5
6  > k <- complete.cases(vec)
7  > k
8  [1] TRUE TRUE TRUE TRUE FALSE TRUE TRUE TRUE↵
      TRUE FALSE
9
10 > vec [k]
11 a1 a2 a3 a4 b1 b2 b3 b4
12  1  2  3  4  3  4  5  6

```

Getting started with Data Frame

- ▶ Data frames are used to store tabular data in R.
- ▶ It is an Important type of object in R, used in a variety of statistical modeling applications.
- ▶ Unlike matrices, data frames can store different classes of objects in each column.
- ▶ Data frames have a special attribute called `row.names` which indicate information about each row of the data frame.

Data Frame

- ▶ Data frames are usually **created by reading in a dataset** using the `read.table(...)` or `read.csv(...)`.
- ▶ Data frames can also be **created explicitly with the `data.frame(...)` function** or they can be **coerced from other types** of objects like lists.
- ▶ Data frames can be **converted to a matrix** by calling `data.matrix(...)`.

```
1 CustomerId,Name,Country,Salary
2 1,AA,MY,$30
3 1,AB,AU,$50
4 1,AC,USA,$60
5 1,AD,FR,$40
6 1,AE,IN,$20
7 1,AF,PK,$15
8
```

Figure 1: CSV format

Creating a Data Frame

Example

```
1 >x <- data.frame(foo = 1:4, bar = c(T, T, F, F))
2 >x
3 foo bar
4 1 TRUE
5 2 TRUE
6 3 FALSE
7 4 FALSE
```

```
1 >nrow(x)
2 [1] 4
3 >ncol(x)
4 [1] 2
```

Creating Data Frames – Explicitly like a list

► Creating data frame

```
1 > measrs <- data.frame(gender = c("M", "M", "F"), ht ←
  = c(172, 186.5, 165), wt = c(91, 99, 74))
2 > measrs
3      gender    ht    wt
4 1      M    172.0   91
5 2      M    186.5   99
6 3      F    165.0   74
```

► Entries in a data.frame are indexed like a matrix:

```
1 measrs[1, 2]
```

Creating Data Frames from Vectors

- ▶ A data frame is used for storing data tables. It is a list of vectors of equal length.

Sample Code

```
1 > n = c(2, 3, 5)
2 > s = c("aa", "bb", "cc")
3 > b = c(TRUE, FALSE, TRUE)
4 > df = data.frame(n, s, b) # df is a data frame
```

Data Frame Attributes

- ▶ Accessing data frame attributes using `names(..)` function.

```
1 > names(measrs)
2 [1] "gender" "ht" "wt"
```

- ▶ Assign names to row.

```
1 > rownames(measrs) <- c("S1", "S2", "S3")
```

- ▶ Access column inside the data frame.

```
1 > measrs$ht
2 [1] 172.0 186.5 165.0
```

Data frame components as Vectors

- ▶ The components of a data frame can be extracted as a vector as in a list:

```
1 > height <- measrs$ht
2 > height
3 [1] 172.0 186.5 165.0
```

- ▶ Character vectors in a data frame are always stored as a **factor**. It's assumed that's what you should do.

```
1 > class(measrs$gend)
2 [1] "factor"
```

Expanding Data Frames

- Components can be added to a data frame in the natural way.

```
1 > measrs$age <- c(28, 55, 43)
2 > measrs
3      gender  ht      wt age
4 S1    M    172.0   91 28
5 S2    M    186.5   99 55
6 S3    F    165.0   74 43
```

Expanding Data Frames

- If you expand the experiment to add data, use row binding to expand.

```

1 > m2 <- data.frame(gender = c("M", "F"), ht = c(
      (170, 166), wt = c(68, 72), age = c(38, 22))
2 > rownames(m2) <- c("S4", "S5")
3 > measrs2 <- rbind(measrs, m2)

```

- If other data are kept on the same samples in another data frame it can be combined with the original using cbind.

Built-in Data Frame

- Lets explore the built-in data frames in R. For example, here is a built-in data frame in R, called mtcars.

mtcars data frame

```
1 > mtcars
2           mpg   cyl  disp    hp  drat
3 Mazda RX4      21.0   6 160.0  110 3.90
4 Mazda RX4 Wag  21.0   6 160.0  110 3.90
5 Datsun 710      22.8   4 108.0   93 3.85
6 Hornet 4 Drive  21.4   6 258.0  110 3.08
7 Hornet Sportabout 18.7   8 360.0  175 3.15
8 Valiant        18.1   6 225.0  105 2.76
9 Duster 360     14.3   8 360.0  245 3.21
```

- Top line of the table called the header contains the column names.
- Horizontal line afterward denotes a data row.

Retrieve data in a cell

- ▶ To retrieve data in a cell, use row and column coordinates in the single square bracket "[]" operator, separated by a comma.
- ▶ The coordinates begins with row position, followed by comma, and ends with the column position.

Sample Code

```
1 > mtcars[1, 2]
2 [1] 6
```

- ▶ We can use the row and column names instead of the numeric coordinates.

```
1 > mtcars["Mazda RX4", "cyl"]
2 [1] 6
```

nrow and ncol functions

- ▶ Lastly, the number of data rows in the data frame is given by the `nrow` function.

Sample code

```
1 > nrow(mtcars)      # number of data rows
2 [1] 32
```

- ▶ And the number of columns of a data frame is given by the `ncol` function.

```
1 > ncol(mtcars)      # number of columns
2 [1] 11
```

Data Frame Column Vector

- ▶ We reference a data frame column with the double square bracket "[[]]" operator.
- ▶ To retrieve the ninth column vector of the built-in data set `mtcars`, we write `mtcars[[9]]`.

Sample Code

```
1 > mtcars[[9]]      # access 9th column
2 [1]  1 1 1 0 0 0 0 0 0 0 0 ...
```

- ▶ We can retrieve the same column vector by its name.

```
1 > mtcars[["am"]]
2 [1]  1 1 1 0 0 0 0 0 0 0 0 ...
```

Data Frame Column Slice

- We retrieve a data frame column slice with the single square bracket "[]" operator.

Sample Code

```
1 > mtcars[1]
2
3 Mazda RX4          mpg
4 Mazda RX4 Wag      21.0
5 Datsun 710          21.0
5 Datsun 710          22.8
```

- We can retrieve the same column slice by its name.

```
1 > mtcars["mpg"]           # access single column
2 > mtcars[c("mpg", "hp")]  # access multiple
                             column ←
```

Data Frame Row Slice

- Retrieve rows from a data frame with the single square bracket operator, in addition to an index vector of row positions, append an extra comma character.

```
1 > mtcars[24,]
2           mpg  cyl  disp  hp drat   wt   ...
3 Camaro Z28 13.3    8   350 245 3.73 3.84   ...
```

- To retrieve more than one rows, use a numeric index vector.

```
1 > mtcars[c(3, 24),] # retrieve multiple row
2 > mtcars["Camaro Z28",] # retrieve a row by its name ←
3 > mtcars[c("Datsun 710", "Camaro Z28"),] # ←
    retrieve multiple row
```

Data Import from CSV File

- ▶ CSV : The data is in comma separated format.
- ▶ The first row represents column names.

```
1 Col1,Col2,Col3
2 100,a1,b1
3 200,a2,b2
4 300,a3,b3
```

- ▶ Code to read from "mydata.csv", using `read.csv(..)`.

```
1 > mydata = read.csv("mydata.csv") # read csv file
2 > mydata
3   Col1 Col2 Col3
4 1   100   a1   b1
5 2   200   a2   b2
6 3   300   a3   b3
```

Working directory

- The data files must be located in the R working directory, which can be found and set with the following functions.

```

1 > getwd()                # get current working ↵
    directory
2 > setwd("<new path>")    # set working directory
3 > setwd("C:/MyDoc")      # Note that the forward ↵
    slash should be used as the path separator even↵
    on Windows platform.
```

Names

- R objects can have names, which is very useful for writing readable code and self-describing objects.

Sample Code

```

1  > x <- 1:3
2  > names(x)
3  NULL
4  > names(x) <- c("New York", "Seattle", "Los Angeles"↵
    ")
5  > x
6  New York  Seattle  Los Angeles
7  1         2         3
8  >names(x)
9  [1] "New York" "Seattle" "Los Angeles"
```


Names

- Lists can also have names, which is often very useful.

Example

```

1  > x <- list("Los Angeles" = 1, Boston = 2, London = 3)
2
3  > x
4  $ Los Angeles
5  [1] 1
6  $ Boston
7  [1] 2
8  $ London
9  [1] 3
10
11 > names(x)
12 [1] "Los Angeles" "Boston" "London"

```

Names

- Matrices can have both column and row names.

Example

```

1 > m <- matrix(1:4, nrow = 2, ncol = 2)
2 > dimnames(m) <- list(c("a", "b"), c("c", "d"))
3 > m
4      c d
5 a 1 3
6 b 2 4

```

Names in Data frame

- ▶ for data frames, there is a separate function for setting the row names, the `row.names()` function.
- ▶ data frames do not have column names, they just have names (like lists).
- ▶ to set the column names of a data frame just use the `names()` function.

Object	Set column names	Set row names
data frame	<code>names()</code>	<code>row.names()</code>
matrix	<code>colnames()</code>	<code>rownames()</code>

Table 1: Quick Summary.

Attributes

- ▶ R objects can have attributes, which are like metadata for the object.
- ▶ metadata can be very useful in that they help to describe the object.
- ▶ For example, column names on a data frame help to tell us what data are contained in each of the columns.

Example

- ▶ names, dimnames
- ▶ dimensions (e.g. matrices, arrays)
- ▶ class (e.g. integer, numeric)
- ▶ length
- ▶ other user-defined attributes/metadata

Reading from Excel

```
1 > install.packages("readxl")
2 > library("readxl")
3 > my_data <- read_excel("C:/myfile.xlsx")

1 #Select file through browsing
2 >my_data <- read_excel(file.choose())

1 # Specify sheet by its name
2 my_data <- read_excel("my_file.xlsx", sheet = "data↵")
```

Summary

There are a variety of different builtin-data types in R. In this section we have reviewed the following.

- ▶ atomic classes: numeric, logical, character, integer, complex
- ▶ vectors, lists
- ▶ factors
- ▶ missing values
- ▶ data frames and matrices

Practise questions

⇒ Create the following data frame, afterwards invert Sex for all individuals.

		Age	Height	Weight	Sex
1					
2	Alex	25	177	57	F
3	Lily	20	165	69	F
4	Mark	21	145	70	M
5	Oliver	24	165	65	M
6	Martha	22	175	71	F

⇒ How many rows and columns does the new data frame have?

⇒ What class of data is in each column?

Exc

⇒ Create this data frame

```
1      Working
2  Alex      YES
3  Lily      NO
4  Mark      YES
5  Oliver    NO
6  Martha    YES
```

- ⇒ Add this data frame column-wise to the previous one.
- ⇒ How many rows and columns does the new data frame have?
- ⇒ What class of data is in each column?

Exc

⇒ Check what class of data is the (built-in data set) `state.center` and convert it to data frame. Hine, see `as.data.frame?`