Lecture 5 Data management: Part II - Reshaping data

This section describes how to prepare data for further analysis. There are situations when we need the **data frame** in a format that is different from the format in which we received it.

- Subsetting data
- Merging data
- aggregate()

5.1 Subset data

Common tasks

- Select/delete columns
- Select/delete rows with or without conditions
- Select columns and rows with or without conditions

Using

- \$ and [,]
- subset() Very powerful!
- dplyr package and the entire tidyverse

Pick your favorite - one is enough.

In [1]:

A data.frame: 4 × 4

names	score	student.no	pass
<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
Lucy	67	student 1	TRUE
John	56	student 2	FALSE
Mark	87	student 3	TRUE
Candy	91	student 4	TRUE

5.1.1 \$ and [,]

Can only pick one variable.

```
In [2]:
```

```
names(df)
```

'names' 'score' 'student.no' 'pass'

In [3]:

```
# Recall the indexing system in R
df$names # Select one variable
```

Lucy John Mark Candy

▶ Levels:

In [4]: # Delete one variable df.copy <- df</pre> df.copy\$names <- NULL</pre> df.copy A data.frame: 4 × 3

score	student.no	pass
<dbl></dbl>	<chr></chr>	<lgl></lgl>
67	student 1	TRUE
56	student 2	FALSE
87	student 3	TRUE
91	student 4	TRUE

In [5]:

```
df[, 2]
```

67 56 87 91

In [6]:

```
df[ , "score"]
```

67 56 87 91

In [7]:

```
str(df[ , "score"]) # 1D vector
```

num [1:4] 67 56 87 91

```
df[ , "score", drop = FALSE]
str(df[ , "score", drop = FALSE]) # 4 x 1 data frame
# The argument "drop = FALSE" maintains the original dimension
# The default is true
Α
data.frame:
4 \times 1
  score
  <dbl>
     67
     56
     87
     91
'data.frame': 4 obs. of 1 variable:
 $ score: num 67 56 87 91
In [9]:
df[1, ]
str(df[1, ]) # 1 x 4 data frame
# Can we drop a dimension here? Why?
A data.frame: 1 × 4
names score student.no pass
 <fct> <dbl> <chr> <lgl>
         67 student 1 TRUE
  Lucy
'data.frame': 1 obs. of 4 variables:
 $ names : Factor w/ 4 levels "Candy", "John",..:
3
         : num 67
 $ score
 $ student.no: chr "student 1"
         : logi TRUE
 $ pass
```

In [8]:

```
In [10]:
```

```
df[1, , drop = TRUE]
```

\$names

Lucy

▶ Levels:

\$score

67

\$student.no

'student 1'

\$pass

TRUE

Any advantage of an n x 1 data frame over a vector of length n? <==> Is the drop argument useful?

```
In [11]:
```

```
# Delete variable "names" + reorder columns
df[ , c("student.no", "score", "pass")]
```

A data.frame: 4 × 3

student.no	score	pass
<chr></chr>	<dbl></dbl>	<lgl></lgl>
student 1	67	TRUE
student 2	56	FALSE
student 3	87	TRUE
student 4	91	TRUE

In [12]:

```
df$pass == TRUE
```

TRUE FALSE TRUE TRUE

In [13]:

```
# Select rows that passed
df[df$pass == TRUE, ]
```

A data.frame: 3 × 4

	names	score	student.no	pass
	<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
1	Lucy	67	student 1	TRUE
3	Mark	87	student 3	TRUE
4	Candy	91	student 4	TRUE

In [14]:

```
df[df$names == "Lucy", ]
```

A data.frame: 1 × 4

```
namesscorestudent.nopass<fct><dbl><chr><lgl>Lucy67student 1TRUE
```

In [15]:

```
# Delete variable
df[ , -c(1, 2)] # Delete the 1st and 2nd
```

```
student.no pass

<chr> <ld><chr> <ld><chr> <ld>student 1 TRUE</ld>

student 2 FALSE

student 3 TRUE

student 4 TRUE
```

```
In [16]:
```

```
# I believe that this used to work, but not anymore.
# df[ , -c("names", "score")]

# Now
drop <- c("names", "score")
df[ , !names(df) %in% drop]</pre>
```

A data.frame: 4 × 2

```
student.nopass<chr><lgI>student 1TRUEstudent 2FALSEstudent 3TRUEstudent 4TRUE
```

In [17]:

```
names(df)
!names(df) %in% drop
```

'names' 'score' 'student.no' 'pass'

FALSE FALSE TRUE TRUE

In [18]:

```
select = c("student.no", "pass")
df[ , names(df) %in% select]
```

A data.frame: 4 × 2

```
student.nopass<chr><lg>>student 1TRUEstudent 2FALSEstudent 3TRUEstudent 4TRUE
```

In [19]:

```
# How does this work?

1 %in% c(1, 3, 5)

"b" %in% c("a", "c", "e")

1:10 %in% c(1, 3, 5)
```

TRUE

FALSE

TRUE FALSE TRUE FALSE FALSE FALSE FALSE

a %in% b checks whether $a \in b$ for every single entry in a.

Exercise: show the name and score of those who passed except Lucy.

In [20]:

```
df[df$pass == TRUE & df$name != "Lucy" , c("names", "score")]
```

A data.frame: 2 × 2

names score

- **3** Mark 87
- **4** Candy 91

In [21]:

```
passed <- df$pass == TRUE
passed</pre>
```

TRUE FALSE TRUE TRUE

In [22]:

```
notLucy <- df$name != "Lucy"
notLucy</pre>
```

FALSE TRUE TRUE TRUE

In [23]:

```
rowCondition <- passed & notLucy rowCondition
```

FALSE FALSE TRUE TRUE

In [24]:

df[rowCondition,]

A data.frame: 2 × 4

	names	score	student.no	pass
	<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
3	Mark	87	student 3	TRUE
4	Candy	91	student 4	TRUE

5.1.2 subset()

```
subset(x, subset, select, drop = FALSE, ...)
```

In [25]:

df

names	score	student.no	pass
<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
Lucy	67	student 1	TRUE
John	56	student 2	FALSE
Mark	87	student 3	TRUE
Candy	91	student 4	TRUE

In [26]:

```
# "select" argument selects columns
subset(df, select = c(student.no, pass))
```

A data.frame: 4 × 2

pass
<lgl></lgl>
TRUE
FALSE
TRUE
TRUE

In [27]:

```
# Can also delete unwanted columns
subset(df, select = -c(names, score))
```

student.no	pass
<chr></chr>	<lgl></lgl>
student 1	TRUE
student 2	FALSE
student 3	TRUE
student 4	TRUE

In [28]:

```
# "subset" argument selects rows
# Can apply conditions
subset(df, subset = (score > 80))
```

A data.frame: 2 × 4

	names	score	student.no	pass
	<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
3	Mark	87	student 3	TRUE
4	Candy	91	student 4	TRUE

In [29]:

```
# Now use both select and subset arguments to apply conditions
# Select the names of those who passed
subset(df, select = names, subset = (pass == TRUE))
```

data.frame:

 3×1

names

	<fct></fct>
1	Lucy

- 3 Mark
- 4 Candy

Note that all subsets are still data frames.

Exercise: show the name and score of those who passed except Lucy.

```
In [30]:
```

```
# Show the name and score of those who passed except Lucy(s).
# Recall logical operators &, | and !
subset(df, select = c(names, score), subset = pass == TRUE & nam
es != "Lucy")
```

A data.frame: 2 × 2

names	score
<fct></fct>	<dbl></dbl>

3 Mark 87

4 Candy 91

5.1.3 dplyr package

'dplyr is a grammar of data manipulation'

https://dplyr.tidyverse.org (https://dplyr.tidyverse.org)

I do not use this package, or any other packages within the whole tidyverse. https://www.tidyverse.org (https://www.tidyverse.org)

```
In [31]:
```

```
library(dplyr)

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':
    filter, lag

The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
```

In [32]:

```
# Show the name and score of those who passed except Lucy(s).
df.col <- filter(df, names != "Lucy" & pass == TRUE)
df.col</pre>
```

A data.frame: 2 × 4

	names	score	student.no	pass
	<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
•	Mark	87	student 3	TRUE
	Candy	91	student 4	TRUE

In [33]:

```
df.final <- select(df.col, c(names, score))
df.final</pre>
```

```
A data.frame: 2 ×
```

2

names score

<fct></fct>	<dbl></dbl>
Mark	87
Candy	91

dplyr cheetsheet

https://rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf (https://rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf)

5.2 Merge data

5.2.1 Add cases/observations to a data frame

This is basically adding rows.

In [34]:

df

A data.frame: 4 × 4

	names	score	student.no	pass
	<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
•	Lucy	67	student 1	TRUE
	John	56	student 2	FALSE
	Mark	87	student 3	TRUE
	Candy	91	student 4	TRUE

In [35]:

names	score	student.no	pass
<fct></fct>	<dbl></dbl>	<fct></fct>	<lgl></lgl>
Name	79	student 5	TRUE
Nom	48	student 6	FALSE

In [36]:

```
df.new <- rbind(df, new.students); df.new</pre>
```

A data.frame: 6 × 4

names	score	student.no	pass
<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
Lucy	67	student 1	TRUE
John	56	student 2	FALSE
Mark	87	student 3	TRUE
Candy	91	student 4	TRUE
Name	79	student 5	TRUE
Nom	48	student 6	FALSE

5.2.2 Add variables to a dataset

This is adding columns.

In [37]:

```
# Option 1
df.copy$id1 <- 1:4
df.copy</pre>
```

score	student.no	pass	id1
<dbl></dbl>	<chr></chr>	<lgl></lgl>	<int></int>
67	student 1	TRUE	1
56	student 2	FALSE	2
87	student 3	TRUE	3
91	student 4	TRUE	4

In [38]:

```
# Option 2
df.copy <- data.frame(df.copy, id2 = 1:4)
df.copy</pre>
```

A data.frame: 4 × 5

score	student.no	pass	id1	id2
<dbl></dbl>	<chr></chr>	<lgl></lgl>	<int></int>	<int></int>
67	student 1	TRUE	1	1
56	student 2	FALSE	2	2
87	student 3	TRUE	3	3
91	student 4	TRUE	4	4

In [39]:

```
# Option 3
id3 <- 1:4
cbind(df.copy, id3)</pre>
```

score	student.no	pass	id1	id2	id3
<dbl></dbl>	<chr></chr>	<lgl></lgl>	<int></int>	<int></int>	<int></int>
67	student 1	TRUE	1	1	1
56	student 2	FALSE	2	2	2
87	student 3	TRUE	3	3	3
91	student 4	TRUE	4	4	4

Easily extend to adding multiple columns.

5.2.3 Merge data frames

Can be very useful when we link databases. For example,

- 1. Database 1 is the electronic health record.
- 2. Database 2 is the claims data for prescription drugs.

We can merge two databases using the unique patient ID.

In [40]:

```
# df stores student's EPIB 613 score df
```

names	score	student.no	pass
<fct></fct>	<dbl></dbl>	<chr></chr>	<lgl></lgl>
Lucy	67	student 1	TRUE
John	56	student 2	FALSE
Mark	87	student 3	TRUE
Candy	91	student 4	TRUE

In [41]:

student.no	major
<fct></fct>	<fct></fct>
student 1	MSc PH
student 2	PhD Epi
student 3	MSc Epi
student 4	MSc PH
student 5	PhD Biostat
student 6	MSc Biostat

In [42]:

```
# See what does the argument 'all' do.
df.full <- merge(df, df.major, by = "student.no", all = TRUE)
df.full</pre>
```

A data.frame: 6 × 5

major	pass	score	names	student.no
<fct></fct>	<lgl></lgl>	<dbl></dbl>	<fct></fct>	<chr></chr>
MSc PH	TRUE	67	Lucy	student 1
PhD Epi	FALSE	56	John	student 2
MSc Epi	TRUE	87	Mark	student 3
MSc PH	TRUE	91	Candy	student 4
PhD Biostat	NA	NA	NA	student 5
MSc Biostat	NA	NA	NA	student 6

5.3 aggregate()

- Very very very useful function!
- It does conditional operations.
 - that requires subsetting when you don't know aggregate()

I need a big and complex dataset.

In [43]:

```
# Some simple simulation
# People who take the drug, that are obese and that are younger
are more likely to be cured.
# Setting seeds make random number generation reproducible.
set.seed(613)
n < -100
drug <- sample(c(0, 1), size = n, replace = TRUE, prob = c(0.8, 1)
0.2))
obesity \leftarrow sample(c(0, 1), size = n, replace = TRUE, prob = c(0).
5, 0.5)
age <- round(rnorm(n, mean = 60, sd = 10))
logit.p <- log(1.8)*drug + log(0.85)*(age - 60) + log(1.2)*obesi
ty + log(0.2)
p <- exp(logit.p)/(1 + exp(logit.p))</pre>
cured <- rbinom(n, size = 1, prob = p)</pre>
sim <- data.frame(drug, obesity, age, cured)</pre>
head(sim, 10)
```

drug	obesity	age	cured
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<int></int>
1	1	53	1
1	1	44	1
0	1	61	1
1	0	41	1
0	0	49	0
1	1	54	1
0	1	51	0
1	1	53	0
0	1	70	0
0	1	54	1

In [44]:

```
# Tabulate exposure and outcome
table(sim[, c("drug", "cured")])
# ~30% among unexposed to the drug are cured
# 40% among exposed are cured
```

```
cured
drug 0 1
0 47 23
1 18 12
```

Quick exercise: calculate the mean age of the exposed group and the unexposed group

In [45]:

```
exposed.group <- sim[sim$drug == 1, ]
head(exposed.group)
mean(exposed.group$age)</pre>
```

A data.frame: 6 × 4

	drug	obesity	age	cured
	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<int></int>
1	1	1	53	1
2	1	1	44	1
4	1	0	41	1
6	1	1	54	1
8	1	1	53	0
21	1	0	69	0

61.0333333333333

```
In [46]:
```

```
exposed.group1 <- subset(x = sim, subset = (drug == 1))
mean(exposed.group1$age)</pre>
```

61.0333333333333

```
In [47]:
```

```
mean(subset(x = sim, subset = (drug == 1))$age)
```

61.0333333333333

```
In [48]:
```

```
mean(subset(x = sim, subset = (drug == 1), select = age, drop =
T))
```

61.0333333333333

Survey

- \$ and [,]2
- subset() 7
- dplyr

aggregate() allows us to aggregate subgroups of the data frame by conditions and then apply a function to all the subgroups.

```
aggregate(x, by, FUN, ...)
```

In [49]:

```
# Syntax 1
aggregate(x = sim$age, by = list(drug.justaname = sim$drug), FUN
= mean)
```

A data.frame: 2 × 2

X	drug.justaname	
<dbl></dbl>	<dbl></dbl>	
59.58571	0	
61.03333	1	

In [50]:

```
# Alternative syntax
# I highly recommend this one
aggregate(age~drug, data = sim, FUN = mean)
```

drug	age	
<dbl></dbl>	<dbl></dbl>	
0	59.58571	
1	61.03333	

In [51]:

```
# Mean age by exposure-obesity group, so 2 binary conditions and
4 subgroups
aggregate(x = sim$age, by = list(drug = sim$drug, obesity = sim$
obesity), FUN = mean)
```

A data.frame: 4 × 3

drug	obesity	X
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
0	0	59.17500
1	0	61.53333
0	1	60.13333
1	1	60.53333

In [52]:

```
aggregate(age~drug+obesity, data = sim, FUN = mean)
```

age	obesity	drug
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
59.17500	0	0
61.53333	0	1
60.13333	1	0
60.53333	1	1

In [53]:

A data.frame: 4 × 4

drug	obesity	V1	V 2
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
0	0	59.17500	0.3500000
1	0	61.53333	0.3333333
0	1	60.13333	0.3000000
1	1	60.53333	0.4666667

In [54]:

```
aggregate(cbind(age, cured)~drug+obesity, data = sim, FUN = mean
)
```

A data.frame: 4 × 4

drug	obesity	age	cured
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
0	0	59.17500	0.3500000
1	0	61.53333	0.3333333
0	1	60.13333	0.3000000
1	1	60.53333	0.4666667

With aggregate(), we are already doing analysis.

Exercise: Get the count in each drug-obesity-cured group using aggregate(

```
In [55]:
table(sim[, c("drug", "obesity", "cured")])

, , cured = 0
    obesity
drug 0 1
    0 26 21
    1 10 8

, , cured = 1
    obesity
drug 0 1
    o 14 9
```

In [56]:

```
aggregate(age~drug+obesity+cured, data = sim, FUN = length)
```

A data.frame: 8 × 4

1 5 7

drug	obesity	cured	age
<dbl></dbl>	<dbl></dbl>	<int></int>	<int></int>
0	0	0	26
1	0	0	10
0	1	0	21
1	1	0	8
0	0	1	14
1	0	1	5
0	1	1	9
1	1	1	7

```
In [57]:
```

```
# What aggregate does? Use print() to print out what R aggregate
s.
aggregate(age~drug, data = sim, FUN = print)
# aggregate the target variables in to vectors, based on conditi
ons.
```

```
[1] 61 49 51 70 54 64 59 73 63 45 65 49 63 68 58 62 49 70 82 59 47 41 64 49 67 [26] 65 71 39 57 53 55 54 62 50 44 49 59 55 58 60 56 47 74 51 76 58 46 63 59 69 [51] 69 59 66 59 63 54 75 80 75 62 59 74 51 74 69 52 52 51 58 57 [1] 53 44 41 54 53 69 81 62 55 77 49 71 76 82 66 68 51 71 54 55 66 65 52 62 70 [26] 47 59 57 50 71
```

A data.frame: 2 × 2

- 0 61, 49, 51, 70, 54, 64, 59, 73, 63, 45, 65, 49, 63, 68, 58, 62, 49, 70, 82, 59, 47, 41, 64, 49, 67, 65, 71, 39, 57, 53, 55, 54, 62, 50, 44, 49, 59, 55, 58, 60, 56, 47, 74, 51, 76, 58, 46, 63, 59, 69, 69, 59, 66, 59, 63, 54, 75, 80, 75, 62, 59, 74, 51, 74, 69, 52, 52, 51, 58, 57
- 53, 44, 41, 54, 53, 69, 81, 62, 55, 77, 49, 71, 76, 82, 66, 68, 51, 71, 54, 55, 66, 65, 52, 62, 70, 47, 59, 57, 50, 71

In [58]:

```
# length() function gives the length of a vector
length(c(1,4,123))
```

3

In []: