Study Guide: Data Manipulation with R

Main concepts

☐ File management – The table below summarizes the useful commands to make sure the working directory is correctly set:

Category	Action	Command	
	Change directory to another path	setwd(path)	
Paths	Get current working directory	getwd()	
	Join paths	file.path(path_1,, path_n)	
	List files and folders in a given directory	list.files(path, include.dirs=TRUE)	
Files	Check if path is a file / folder	file_test('-f', path)	
	Check if path is a file / folder	file_test('-d', path)	
	Read / write csv file	read.csv(path_to_csv_file)	
	nead / write csv ille	write.csv(df, path_to_csv_file)	

☐ Chaining — The symbol %>%, also called "pipe", enables to have chained operations and provides better legibility. Here are its different interpretations:

- f(arg_1, arg_2, ..., arg_n) is equivalent to arg_1%>%f(arg_2, arg_3, ..., arg_n), and also to:
 - arg_1%>%f(., arg_2, ..., arg_n)
 - arg_2%>%f(arg_1, ., arg_3, ..., arg_n)
 - arg_n%>%f(arg_1, ..., arg_n-1, .)
- A common use of pipe is when a dataframe df gets first modified by some operation 1, then some operation 2, until some operation in a sequential way. It is done as follows:

df gets some_operation_1, then some_operation_2, ..., # then df%>%
some_operation_1
some_operation_2
...
some_operation_n
some_operation_n

☐ Exploring the data – The table below summarizes the main functions used to get a complete overview of the data:

Category	Action	Command	
	Select columns of interest	df%>%select(col_list)	
Look at data Remove unwanted columns df%>%select(-col_list)		df%>%select(-col_list)	
200k at data	Look at n first rows / last rows	df%>%head(n) / df%>%tail(n)	
	Summary statistics of columns	df%>%summary()	
Data types	Data types of columns	df%>%str()	
Data types	Number of rows / columns	df%>%NROW() / df%>%NCOL()	

☐ Data types – The table below sums up the main data types that can be contained in columns:

Data type	Description	Example
character	String-related data	'teddy bear'
factor	String-related data that can be put in bucket, or ordered	'high'
numeric	Numerical data	24.0
int	Numeric data that are integer	24
Date	Dates	'2020-01-01'
POSIXct	Timestamps	'2020-01-01 00:01:00'

Data preprocessing

☐ **Filtering** – We can filter rows according to some conditions as follows:

	Category	Operation	Command
df%>%		Equality / non-equality	== / !=
filter(son	Basic ne_col some_op	eration some_value_or_list_or	<, <=, >=, >
where some_o	peration is one	of And following:	& /
		Check for missing value	is.na()
	Advanced	Belonging	%in%(val_1,, val_n)
		Pattern matching	%like%'val'

Remark: we can filter columns with the select_if command.

☐ Changing columns – The table below summarizes the main columnoperations:

Action	Command
Add new columns on top of old ones	df%>%mutate(new_col=operation(other_cols))
Add new columns and discard old ones	df%>%transmute(new_col=operation(other_cols))
Modify several columns in- place	df%>%mutate_at(vars, funs)
Modify all columns in-place	df%>%mutate_all(funs)
Modify columns fitting a specific condition	df%>%mutate_if(condition, funs)
Unite columns	df%>%unite(new_merged_col, old_cols_list)
Separate columns	df%>%separate(col_to_separate, new_cols_list)

☐ Conditional column — A column can take different values with respect to a particular set of conditions with the case_when() command asfollows:

 $\begin{array}{c} {\sf case_when(condition_1 \sim value_1, \#\, lf\, condition_1\, then\, value_1}\\ {\sf condition_2 \sim value_2, \#\, lf\, condition_2\, then\, value_2} \end{array}$

TRUE ~ value n) # C

Otherwise, value_n

Remark: the ifelse(condition_if_true, value_true, value_other) can be used and is easier to manipulate if there is only one condition.

 $\hfill \square$ Mathematical operations — The table below sums up the main mathematical operations that can be performed on columns:

Operation	Command	
$\sqrt{\bar{x}}$	sqrt(x)	
x∫	floor(x)	
x	ceiling(x)	

☐ **Datetime conversion** – Fields containing datetime values can be stored in two different POSIXt data types:

Action	Command
Converts to datetime with seconds since origin	as.POSIXct(col, format)
Converts to datetime with attributes (e.g. time zone)	as.POSIXIt(col, format)

where format is a string describing the structure of the field and using the commands summarized in the table below:

Category	Command	Description	Example
Year	′%Y′ / ′%y′	With / without century	2020 / 20
Month	'%B' / '%b' / '%m'	Full / abbreviated / numerical	August / Aug/ 8
Weekday	'%A' / '%a'	Full / abbreviated	Sunday / Sun
, ,	'%u' / '%w'	Number (1-7) / Number (0-6)	7/0
Day	'%d' / '%j'	Of the month / of the year	09 / 222
Time	'%H' / '%M'	Hour / minute	09 / 40
Timezone	′%Z′ / ′%z′	String / Number of hours from UTC	EST / -0400

Remark: data frames only accept datetime in POSIXct format.

 $\hfill\Box$ Date properties – In order to extract a date-related property from a date time object, the following command is used:



format(datetime_object, format)

where format follows the same convention as in the table above.

Data frame transformation

☐ Merging data frames – We can merge two data frames by a given field as follows:



merge(df_1, df_2, join_field, join_type)

where join field indicates fields where the join needs to happen:

Case	Fields are equal	Different field names
Command	by='field'	by.x='field_1', by.y='field_2'

and where join_type indicates the join type, and is one of the following:

Join type	Option	Illustration
Inner join	default	df_1 df_2
Left join	all.x=TRUE	df_1 df_2
Right join	all.y=TRUE	df_1 df_2
Full join	all=TRUE	df_1 df_2

Remark: if the by parameter is not specified, the merge will be a cross join.

☐ Concatenation – The table below summarizes the different ways data frames can be con- catenated:

Туре	Command	Illustration
		df_1
Rows	rbind(df_1,, df_n)	df_2
		:
		df_n
Columns	cbind(df_1,, df_n)	
		df_1 df_2 df_n

☐ **Common transformations** – The common data frame transformations are summarized in the table below:

Туре	Command	Illustration	
,,		Before	After
Long to wide	spread(df, key='key', value='value'	some_cols key value	some_cols key_1 key_n
Wide to long	gather(df, key='key' value='value', c(key_1,, key_n)	some_cols key_1 key_n	some_cols key value

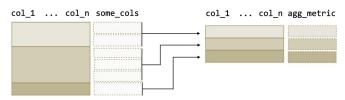
☐ **Row operations** – The following actions are used to make operations on rows of the data frame:

Action	Command	Illustration	
		Before	After
Sort with respect to columns	df%>% arrange(col_1,, col_n)	col_1 col_2 col_3 other_cols	col_1 col_2 col_3 other_cols
Dropping duplicates	df%>%unique()	col_1 col_2 col_3 col_4 col_5	col_1 col_2 col_3 col_4 col_5
Drop rows with at least a null value	df%>%na.omit()	col_1 col_2 col_3 col_4 col_5	col_1 col_2 col_3 col_4 col_5

Remark: by default, the arrange command sorts in ascending order. If we want to sort it in descending order, the - command needs to be used before a column.

Aggregations

☐ **Grouping data** – Aggregate metrics are computed across groups as follows:



The R command is as follows:

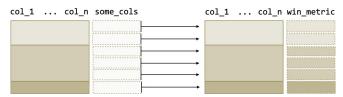


 $\hfill\Box$ Aggregate functions— The table below summarizes the main aggregate functions that can be used in an aggregation query:

Category	Action	Command
Properties	Count of observations	n()
Values	Sum of values of observations	sum()
	Max / min of values of observations	max() / min()
	Mean / median of values of observations	mean() / median()
	Standard deviation / variance across observations	sd() / var()

Window functions

☐ **Definition** – A window function computes a metric over groups and has the following structure:



The R command is as follows:



Remark: applying a window function will not change the initial number of rows of the data frame.

□ Row numbering – The table below summarizes the main commands that rank each row across specified groups, ordered by a specific field:

Join type	Command	Example
row_number(x)	Ties are given different ranks	1, 2, 3, 4
rank(x)	Ties are given same rank and skip numbers	1, 2.5, 2.5, 4
dense_rank(x)	Ties are given same rank and do not skip numbers	1, 2, 2, 3

□ Values – The following window functions allow to keep track of specific types of values with respect to the group:

Command	Description
first(x)	Takes the first value of the column
last(x)	Takes the last value of the column
lag(x, n)	Takes the $n^{ m th}$ previous value of the column
lead(x, n)	Takes the $n^{ m th}$ following value of the column
nth(x, n)	Takes the $n^{ m th}$ value of the column