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SECTION 1

### Data retrieval with SQL

### 1.1 General concepts

□ Structured Query Language – Structured Query Language, abbreviated as SQL, is a language that is largely used in the industry to query data from databases.

□ Query structure – Queries are usually structured as follows:

```
\mathbf{SQL}
-- Select fields
                                      mandatory
SELECT
    col_1,
    col_2,
    col_n
-- Source of data
                                      mandatory
FROM table t
-- Gather info from other sources
                                      optional
JOIN other_table ot
 ON (t.key = ot.key)
-- Conditions
                                      optional
WHERE some_condition(s)
-- Aggregating
                                      optional
GROUP BY column_group_list
-- Sorting values
                                      optional
ORDER BY column_order_list
-- Restricting aggregated values
                                      optional
HAVING some_condition(s)
-- Limiting number of rows
                                      optional
LIMIT some_value
```

Remark: the SELECT DISTINCT command can be used to ensure not having duplicate rows.

☐ Condition – A condition is of the following format:

```
SQL
some_col some_operator some_col_or_value
```

where some\_operator can be among the following common operations:

Category	Operator	Command
	Equality / non-equality	= / !=, <>
	Inequalities	>=, >, <, <=
General	Belonging	IN (val_1,, val_n)
General	And / or	AND / OR
	Check for missing value	IS NULL
	Between bounds	BETWEEN val_1 AND val_2
Strings	Pattern matching	LIKE '%val%'

□ Joins – Two tables table\_1 and table\_2 can be joined in the following way:

```
SQL
...

FROM table_1 t1
type_of_join table_2 t2
ON (t2.key = t1.key)
...
```

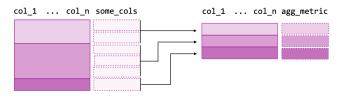
where the different type\_of\_join commands are summarized in the table below:

Type of join	Illustration
INNER JOIN	t1 t2
LEFT JOIN	t1 t2
RIGHT JOIN	t1 t2
FULL JOIN	t1 t2

Remark: joining every row of table 1 with every row of table 2 can be done with the CROSS JOIN command, and is commonly known as the cartesian product.

# 1.2 Aggregations

 $\hfill\Box$  Grouping data – Aggregate metrics are computed on grouped data in the following way:



The SQL command is as follows:

```
SQL

SELECT

col_1,
agg_function(col_2)

FROM table
GROUP BY col_1
```

□ Grouping sets – The GROUPING SETS command is useful when there is a need to compute aggregations across different dimensions at a time. Below is an example of how all aggregations across two dimensions are computed:

```
SQL

SELECT

col_1,
col_2,
agg_function(col_3)

FROM table

GROUP BY (
GROUPING SETS
(col_1),
(col_2),
(col_2),
(col_1, col_2)
)
```

 $\square$  Aggregation functions – The table below summarizes the main aggregate functions that can be used in an aggregation query:

Category	Operation	Command
	Mean	AVG(col)
	Percentile	PERCENTILE_APPROX(col, p)
Values	Sum / # of instances	SUM(col) / COUNT(col)
	Max / min	MAX(col) / MIN(col)
	Variance / standard deviation	VAR(col) / STDEV(col)
Arrays	Concatenate into array	collect_list(col)

Remark: the median can be computed using the PERCENTILE\_APPROX function with p equal to 0.5.

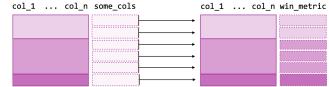
 $\square$  Filtering – The table below highlights the differences between the WHERE and HAVING commands:

WHERE	HAVING	
- Filter condition applies to individual rows - Statement placed right after FROM	- Filter condition applies to aggregates - Statement placed right after GROUP BY	

Remark: if WHERE and HAVING are both in the same query, WHERE will be executed first.

#### 1.3 Window functions

 $\hfill\Box$  Definition – A window function computes a metric over groups and has the following structure:



The SQL command is as follows:

```
SQL some_window_function() OVER(PARTITION BY some_col ORDER BY another_col)
```

Remark: window functions are only allowed in the SELECT clause.

 $\square$  Row numbering – The table below summarizes the main commands that rank each row across specified groups, ordered by a specific column:

Command	Description	Example
ROW_NUMBER()	Ties are given different ranks	1, 2, 3, 4
RANK()	Ties are given same rank and skip numbers	1, 2, 2, 4
DENSE_RANK()	Ties are given same rank and don't skip numbers	1, 2, 2, 3

 $\square$  Values – The following window functions allow to keep track of specific types of values with respect to the partition:

Command	Description
FIRST_VALUE(col)	Takes the first value of the column
LAST_VALUE(col)	Takes the last value of the column
LAG(col, n)	Takes the $n^{\mathrm{th}}$ previous value of the column
LEAD(col, n)	Takes the $n^{\mathrm{th}}$ following value of the column
NTH_VALUE(col, n)	Takes the $n^{\text{th}}$ value of the column

#### 1.4 Advanced functions

 $\hfill\Box$  SQL tips – In order to keep the query in a clear and concise format, the following tricks are often done:

Operation	Command	Description
Renaming columns	SELECT operation_on_column AS col_name	New column names shown in query results
Abbreviating tables	FROM table_1 t1	Abbreviation used within query for simplicity in notations
Simplifying group by	GROUP BY col_number_list	Specify column position in SELECT clause instead of whole column names
Limiting results	LIMIT n	Display only n rows

 $\square$  Sorting values – The query results can be sorted along a given set of columns using the following command:

```
SQL
... [query] ...
ORDER BY col_list
```

Remark: by default, the command sorts in ascending order. If we want to sort it in descending order, the DESC command needs to be used after the column.

 $\square$  Column types – In order to ensure that a column or value is of one specific data type, the following command is used:

```
CAST(some_col_or_value AS data_type)
```

where data\_type is one of the following:

Data type	Description	Example
INT	Integer	2
DOUBLE	Numerical value	2.0
STRING	String	'teddy bear'
VARCHAR		
DATE	Date	'2020-01-01'
TIMESTAMP	Timestamp	'2020-01-01 00:00:00.000'

Remark: if the column contains data of different types, the  $TRY\_CAST()$  command will convert unknown types to NULL instead of throwing an error.

□ Column manipulation – The main functions used to manipulate columns are described in the table below:

Category	Operation	Command
Take first non-NULL value		COALESCE(col_1, col_2,, col_n)
General	Create a new column combining existing ones	CONCAT(col_1,, col_n)
Value	Round value to n decimals	ROUND(col, n)
	Converts string column to lower / upper case	LOWER(col) / UPPER(col)
	Replace occurrences of old in col to new	REPLACE(col, old, new)
String	Take the substring of col, with a given start and length	SUBSTR(col, start, length)
	Remove spaces from the left / right / both sides	LTRIM(col) / RTRIM(col) / TRIM(col)
	Length of the string	LENGTH(col)
Date	Truncate at a given granularity (year, month, week)	DATE_TRUNC(time_dimension, col_date)
	Transform date	DATE_ADD(col_date, number_of_days)

 $\square$  Conditional column – A column can take different values with respect to a particular set of conditions with the CASE WHEN command as follows:

```
CASE WHEN some_condition THEN some_value
...
WHEN some_other_condition THEN some_other_value
ELSE some_other_value_n END
```

 $\square$  Combining results – The table below summarizes the main ways to combine results in queries:

Category	Command	Remarks
	UNION	Guarantees distinct rows
Union	UNION ALL	Potential newly-formed duplicates are kept
Intersection	INTERSECT	Keeps observations that are in all selected queries

 $\square$  Common table expression – A common way of handling complex queries is to have temporary result sets coming from intermediary queries, which are called common table expressions (abbreviated CTE), that increase the readability of the overall query. It is done thanks to the WITH ... AS ... command as follows:

```
WITH cte_1 AS (
SELECT ...
),
```

```
cte_n AS (
SELECT ...
)

SELECT ...
FROM ...
```

## 1.5 Table manipulation

☐ **Table creation** – The creation of a table is done as follows:

where [table\_type], [creation\_type] and [options] are one of the following:

Category	Command	Description
	Blank	Default table
Table type	EXTERNAL TABLE	External table
Creation type	Blank	Creates table and overwrites current one if it exists
	IF NOT EXISTS	Only creates table if it does not exist
Options location 'path_to_hdfs_folder		Populate table with data from hdfs folder
	stored as data_format	Stores the table in a specific data format, e.g. parquet, orc or avro

 $\square$  Data insertion – New data can either append or overwrite already existing data in a given table as follows:

```
SQL
WITH ... -- optional
INSERT [insert_type] table_name -- mandatory
SELECT ...; -- mandatory
```

where [insert\_type] is among the following:

Command	Description
OVERWRITE	Overwrites existing data
INTO	Appends to existing data

□ **Dropping table** – Tables are dropped in the following way:

```
SQL
DROP TABLE table_name;
```

 $\square$  View – Instead of using a complicated query, the latter can be saved as a view which can then be used to get the data. A view is created with the following command:

```
SQL
CREATE VIEW view_name AS complicated_query;
```

Remark: a view does not create any physical table and is instead seen as a shortcut.

SECTION 2

### Working with data with R

### 2.1 Data manipulation

### 2.1.1 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	<pre>file.path(path_1,, path_n)</pre>
	List files and folders in a given directory	<pre>list.files(path, include.dirs = TRUE)</pre>
Eiles	Files Check if path is a file / folder	<pre>file_test('-f', path)</pre>
riies		<pre>file_test('-d', path)</pre>
Read / write csv file	5 1/ 11 21	read.csv(path_to_csv_file)
	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\_n in a sequential way. It is done as follows:

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

some_operation_1 %>%
some_operation_2 %>%
... %>%
some_operation_n
```

 $\square$  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)
	Remove unwanted columns	df %>% select(-col_list)
Look at data	Look at $n$ first rows / last rows	df %>% head(n) / df %>% tail(n)
	Summary statistics of columns	df %>% summary()
D	Data types of columns	df %>% str()
Data types	Number of rows / columns	df %>% NROW() / df %>% NCOL()

 $\square$  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'

# 2.1.2 Data preprocessing

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

```
df %>%
  filter(some_col some_operation some_value_or_list_or_col)
```

where some\_operation is one of the following:

Category	Operation	Command
	Equality / non-equality	== / !=
Basic	Inequalities	<, <=, >=, >
	And / or	& /
	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 column operations:

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

 $\square$  Conditional column – A column can take different values with respect to a particular set of conditions with the case\_when() command as follows:

```
R

case_when(condition_1 ~ value_1,  # If condition_1 then value_1 condition_2 ~ value_2,  # If condition_2 then value_2 ...

TRUE ~ value_n)  # 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.

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

Operation	Command
$\sqrt{x}$	sqrt(x)
$\lfloor x \rfloor$	floor(x)
$\lceil x \rceil$	ceiling(x)

 $\hfill \Box$  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.POSIX1t(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
XX711	'%A' / '%a'	Full / abbreviated	Sunday / Sun
Weekday	'%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.

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

```
R
format(datetime_object, format)
```

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

## 2.1.3 Data frame transformation

 $\hfill \square$  Merging data frames – We can merge two data frames by a given field as follows:

```
R
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'	<pre>by.x = 'field_1', by.y = 'field_2'</pre>

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
Full join	all = TRUE	df_1 df_2

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

 $\hfill\Box$  Concatenation – The table below summarizes the different ways data frames can be concatenated:

Type	Command	Illustration
Rows	rbind(df_1,, df_n)	df_1  df_2  : df_n
Columns	<pre>cbind(df_1,, df_n)</pre>	df_1 df_2 df_n

 $\hfill\Box$  Common transformations – The common data frame transformations are summarized in the table below:

Type	Command	Illustration	
	Communa	Before	After
Long to wide	<pre>spread(   df, key = 'key',   value = 'value' )</pre>	some_cols key value   key_1	some_cols key_1 key_n
Wide to long	<pre>gather(   df, key = 'key'   value = 'value',   c(key_1,, key_n) )</pre>	some_cols key_1 key_n	some_cols key value

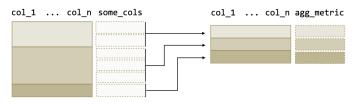
 $\square$  Row operations – The following actions are used to make operations on rows of the data frame:

Action	Command	Illustration	
		Before	After
Sort with respect to columns	<pre>df %&gt;%    arrange(col_1,, col_n)</pre>	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.

# 2.1.4 Aggregations

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



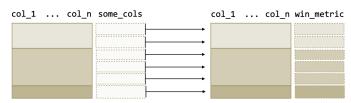
The R command is as follows:

 $\square$  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()
	Sum of values of observations	sum()
Values	Max / min of values of observations	max() / min()
	Mean / median of values of observations	mean() / median()
	Standard deviation / variance across observations	sd() / var()

#### 2.1.5 Window functions

 $\hfill\Box$  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 Exam	
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 $n^{\text{th}}$ previous value of the column
lead(x, n)	Takes the $n^{\text{th}}$ following value of the column
nth(x, n)	Takes the $n^{\text{th}}$ value of the column

### 2.2 Data visualization

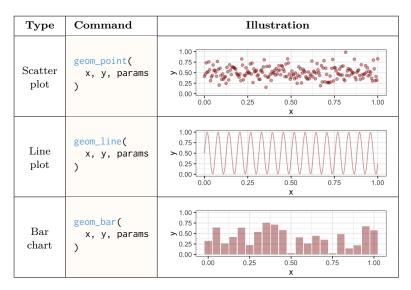
### 2.2.1 General structure

□ Overview – The general structure of the code that is used to plot figures is as follows:

```
ggplot(...) +  # Initialization
geom_function(...) +  # Main plot(s)
facet_function(...) +  # Facets (optional)
labs(...) +  # Legend (optional)
scale_function(...) +  # Scales (optional)
theme_function(...) # Theme (optional)
```

We note the following points:

- The ggplot() layer is mandatory.
- When the data argument is specified inside the ggplot() function, it is used as default in the following layers that compose the plot command, unless otherwise specified.
- In order for features of a data frame to be used in a plot, they need to be specified inside the <code>aes()</code> function.
- ☐ Basic plots The main basic plots are summarized in the table below:

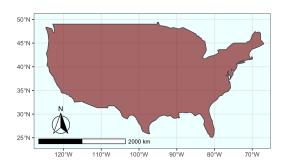


Type	Command	Illustration
Box plot	<pre>geom_boxplot(    x, y, params )</pre>	1.00 0.75 > 0.50 0.25 0.00 group_1 group_2 group_3 group_4 group_5
Heatmap	<pre>geom_tile(    x, y, params )</pre>	group_5 group_4 group_3 group_2 group_1 0.00 0.25 0.50 0.75 1.00

where the possible parameters are summarized in the table below:

Command	Description Use case	
color	Color of a line / point / border	'red'
fill	Color of an area	'red'
size	Size of a line / point	4
shape	Shape of a point	4
linetype	Shape of a line	'dashed'
alpha	Transparency, between 0 and 1	0.3

☐ Maps – It is possible to plot maps based on geometrical shapes as follows:



The following table summarizes the main commands used to plot maps:

Category	Action	Command
Map	Draw polygon shapes from the geometry column	geom_sf(data)
Additional	Add and customize geographical directions	annotation_north_arrow(l)
elements	Add and customize distance scale	annotation_scale(1)
Range	Customize range of coordinates	<pre>coord_sf(xlim, ylim)</pre>

□ Animations – Plotting animations can be made using the gganimate library. The following command gives the general structure of the code:

```
# Main plot
ggplot() +
    ... +
    transition_states(field, states_length)
# Generate and save animation
animate(plot, duration, fps, width, height, units, res, renderer)
anim_save(filename)
```

### 2.2.2 Advanced features

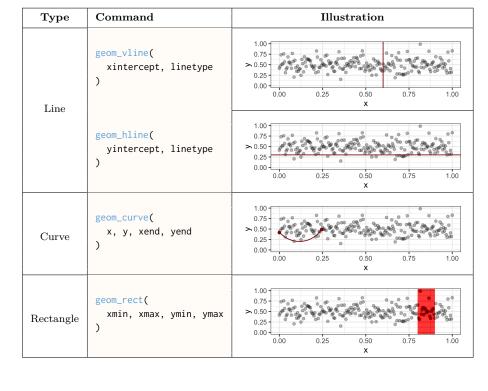
 $\hfill\Box$  Facets – It is possible to represent the data through multiple dimensions with facets using the following commands:

Type	Command	Illustration
Grid (1 or 2D)	<pre>facet_grid(   row_var ~ column_var )</pre>	group_1 group_2 group_3 group_4 0.75
Wrapped	<pre>facet_wrap(   vars(x1,, xn),   nrow, ncol )</pre>	group_1 group_2 group_2 group_3 group_4 group_4 your property of the property

☐ Text annotation – Plots can have text annotations with the following commands:

Command	Illustration
<pre>geom_text(   x, y, label,   hjust, vjust )</pre>	1.00 0.75 > 0.50 0.25 0.00 0.00 0.25 0.00 0.25 0.00 0.25 0.00 0.25
<pre>geom_label_repel(    x, y, label,    nudge_x, nudge_y )</pre>	0.75 > 0.50 0.25 0.00 0.00 0.25 0.00 0.25 0.00 0.25 0.00 0.25

☐ Additional elements – We can add objects on the plot with the following commands:



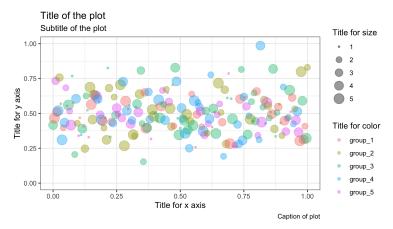
## 2.2.3 Last touch

 $\hfill\Box$  Legend – The title of legends can be customized to the plot with the following command:

where the params are summarized below:

Element	Command
Title / subtitle of the plot	<pre>title = 'text' / subtitle = 'text'</pre>
Title of the $x / y$ axis	x = 'text' / y = 'text'
Title of the size / color	size = 'text' / color = 'text'
Caption of the plot	caption = 'text'

This results in the following plot:



 $\Box$  Plot appearance – The appearance of a given plot can be set by adding the following command:

Type	Command	Illustration
Black and white	theme_bw()	100 0.75 > 0.50 0.00 0.00 0.25 0.50 0.50 0.75 100
Classic	theme_classic()	100 0.75 > 0.50 0.00 0.00 0.25 0.50 0.50 0.75 1.00
Minimal	theme_minimal()	1.00 0.75 > 0.50 0.00
None	theme_void()	

In addition, theme() is able to adjust positions/fonts of elements of the legend.

Remark: in order to fix the same appearance parameters for all plots, the theme\_set() function can be used.

 $\square$  Scales and axes – Scales and axes can be changed with the following commands:

Category	Action	Command
	Specify range of x / y axis	xlim(xmin, xmax)
Range	Specify range of x / y axis	ylim(ymin, ymax)
		scale_x_continuous()
Nature	Display ticks in a customized manner	scale_x_discrete()
		scale_x_date()
		scale_x_log10()
Magnitude	Transform axes	scale_x_reverse()
		<pre>scale_x_sqrt()</pre>

Remark: the  $scale_x()$  functions are for the x axis. The same adjustments are available for the y axis with  $scale_y()$  functions.

□ Double axes – A plot can have more than one axis with the sec.axis option within a given scale function scale\_function(). It is done as follows:

```
scale_function(sec.axis = sec_axis(~ .))
```

□ Saving figure – It is possible to save figures with predefined parameters regarding the scale, width and height of the output image with the following command:

```
ggsave(plot, filename, scale, width, height)
```

SECTION 3

### Working with data with Python

### 3.1 Data manipulation

## 3.1.1 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	os.chdir(path)	
Paths	Get current working directory	os.getcwd()	
	Join paths	os.path.join(path_1,, path_n)	
	List files and folders in a directory	os.listdir(path)	
Files		os.path.isfile(path)	
	Check if path is a file / folder	os.path.isdir(path)	
	D 1/ Cl	<pre>pd.read_csv(path_to_csv_file)</pre>	
	Read / write csv file	<pre>df.to_csv(path_to_csv_file)</pre>	

 $\Box$  Chaining – It is common to have successive methods applied to a data frame to improve readability and make the processing steps more concise. The method chaining is done as follows:

 $\square$  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[col_list]	
Look at data	Remove unwanted columns	<pre>df.drop(col_list, axis=1)</pre>	
	Look at $n$ first rows / last rows	<pre>df.head(n) / df.tail(n)</pre>	
Summary statistics of columns		df.describe()	
D 41	Data types of columns	df.dtypes / df.info()	
Paths	Number of (rows, columns)	df.shape	

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

Data type	Description	Example
object	String-related data	'teddy bear'
float64	Numerical data	24.0
int64	Numeric data that are integer	24
datetime64	Timestamps	'2020-01-01 00:01:00'

# 3.1.2 Data preprocessing

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

```
Python

df[df['some_col'] some_operation some_value_or_list_or_col]
```

where some\_operation is one of the following:

Category	Operation	Command	
	Equality / non-equality	== / !=	
Basic	Inequalities	<, <=, >=, >	
	And / or	& /	
	Check for missing value	pd.isnull()	
Advanced	Belonging	.isin([val_1,, val_n])	
	Pattern matching	.str.contains('val')	

□ Changing columns – The table below summarizes the main column operations:

Operation	Command	
Add new columns on top of old ones	<pre>df.assign(   new_col=lambda x: some_operation(x) )</pre>	
Rename columns	<pre>df.rename(columns={    'current_col': 'new_col_name'}) })</pre>	
Unite columns	<pre>df['new_merged_col'] = (    df[old_cols_list].agg('-'.join, axis=1) )</pre>	

 $\square$  Conditional column – A column can take different values with respect to a particular set of conditions with the np.select() command as follows:

```
Python
np.select(
  [condition_1, ..., condition_n],  # If condition_1, ..., condition_n
  [value_1, ..., value_n],  # Then value_1, ..., value_n respectively
  default=default_value  # Otherwise, default_value
)
```

Remark: the np.where(condition\_if\_true, value\_true, value\_other) command can be used and is easier to manipulate if there is only one condition.

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

Operation	Command
$\sqrt{x}$	np.sqrt(x)
$\lfloor x \rfloor$	np.floor(x)
$\lceil x \rceil$	np.ceil(x)

□ Datetime conversion – Fields containing datetime values are converted from string to datetime as follows:

```
Python
pd.to_datetime(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
XX711	'%A' / '%a'	Full / abbreviated	Sunday / Sun
Weekday	'%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

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

```
Python
datetime_object.strftime(format)
```

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

#### 3.1.3 Data frame transformation

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

```
Python
df1.merge(df2, join_field, join_type)
```

where join\_field indicates fields where the join needs to happen:

Case	Fields are equal	Fields are different
Command	on='field'	<pre>left_on='field_1', right_on='field_2'</pre>

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

Join type	Option	Illustration	
Inner join	how='inner'	df_1 df_2	
Left join	how='left'	df_1 df_2	
Right join	how='right'	df_1 df_2	
Full join	how='outer'	df_1 df_2	

Remark: a cross join can be done by joining on an undifferentiated column, typically done by creating a temporary column equal to 1.

 $\hfill\Box$  Concatenation – The table below summarizes the different ways data frames can be concatenated:

Type	Command	Illustration
Rows	<pre>pd.concat([df_1,, df_n], axis=0)</pre>	df_1  df_2  : df_n
Columns	<pre>pd.concat([df_1,, df_n], axis=1)</pre>	df_1 df_2 df_n

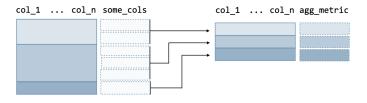
 $\hfill\Box$  Common transformations – The common data frame transformations are summarized in the table below:

Type	Command	Illustration	
-JP0		Before	After
Long to wide	<pre>pd.pivot_table(    df, values='value',    index=some_cols,    columns='key',    aggfunc=np.sum )</pre>	key_1	some_cols key_1 key_n
Wide to long	<pre>pd.melt(     df, var_name='key',     value_name='value',     value_vars=[         'key_1',, 'key_n'     ], id_vars=some_cols )</pre>	some_cols key_1 key_n	key_n

Action	Command	Illustration	
Trough		Before	After
Sort with respect to columns	<pre>df.sort_values(    by=['col_1',, 'col_n'],    ascending=True )</pre>	col_1 col_2 col_3 other_cols	col_1 col_2 col_3 other_cols
Dropping duplicates	<pre>df.drop_duplicates()</pre>	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.dropna()	col_1 col_2 col_3 col_4 col_5	col_1 col_2 col_3 col_4 col_5

# 3.1.4 Aggregations

☐ Grouping data – A data frame can be aggregated with respect to given columns as follows:



The Python command is as follows:

```
Python
.groupby(['col_1', ..., 'col_n'])
.agg({'col': builtin_agg})
```

□ Row operations - The following actions are used to make operations on rows of the data where builtin\_agg is among the following: frame:

Category	Action	Command
Properties	Count of observations	'count'
	Sum of values of observations	'sum'
Values	Max / min of values of observations	'max' / 'min'
varues	Mean / median of values of observations	'mean' / 'median'
	Standard deviation / variance across observations	'std' / 'var'

 $\square$  Custom aggregations – It is possible to perform customized aggregations by using lambda functions as follows:

```
Python

df_agg = (
    df
        .groupby(['col_1', ..., 'col_n'])
        .apply(lambda x: pd.Series({
            'agg_metric': some_aggregation(x)
        })))
)
```

```
Join type Command Example

x.rank(method='first') Ties are given different ranks 1, 2, 3, 4

x.rank(method='min') Ties are given same rank and skip numbers 1, 2.5, 2.5, 4

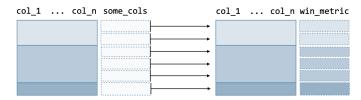
x.rank(method='dense') Ties are given same rank and do not skip numbers 1, 2, 2, 3
```

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

Command	Description
x.shift(n)	Takes the $n^{\rm th}$ previous value of the column
x.shift(-n)	Takes the $n^{\text{th}}$ following value of the column

#### 3.1.5 Window functions

 $\hfill\Box$  Definition – A window function computes a metric over groups and has the following structure:



The Python 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:

### 3.2 Data visualization

### 3.2.1 General structure

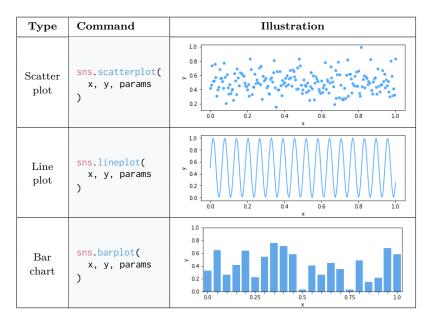
□ Overview – The general structure of the code that is used to plot figures is as follows:

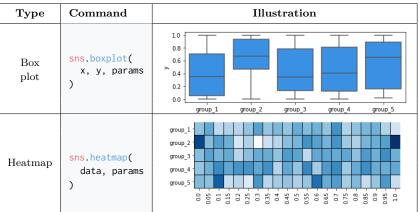
```
# Plot
f, ax = plt.subplots(...)
ax = sns...

# Legend
plt.title()
plt.xlabel()
plt.ylabel()
```

We note that the plt.subplots() command enables to specify the figure size.

☐ Basic plots – The main basic plots are summarized in the table below:





where the meaning of parameters are summarized in the table below:

Command	Description	Use case
hue	Color of a line / point / border	'red'
fill	Color of an area	'red'
size	Size of a line / point	4
linetype	Shape of a line	'dashed'
alpha	Transparency, between 0 and 1	0.3

## 3.2.2 Advanced features

☐ **Text annotation** – Plots can have text annotations with the following commands:

Type	Command	Illustration	
Text	<pre>ax.text(   x, y, s, color )</pre>	10 08 point 1  > 0.6 0.4 0.2 0.4 0.6 0.8 1.0	

□ Additional elements – We can add objects on the plot with the following commands:

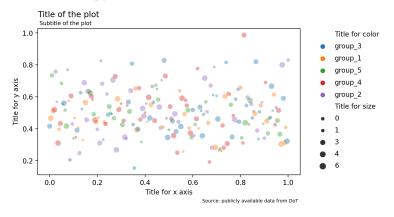
Type	Command	Illustration	
Line	<pre>ax.axvline(   x, ymin, ymax, color,   linewidth, linestyle )</pre>	10 08 > 06 04 02 00 02 04 06 08 10	
	<pre>ax.axhline(   y, xmin, xmax, color,   linewidth, linestyle )</pre>	10 08 > 06 04 02 00 0'2 0'4 0'6 0'8 1'0	
Rectangle	<pre>ax.axvspan(   xmin, xmax, ymin, ymax,   color, fill, alpha )</pre>	10 08 > 06 04 02 00 02 04 06 08 10	

# 3.2.3 Last touch

 $\hfill\Box$  Legend – The title of legends can be customized to the plot with the commands summarized below:

Element	Command	
TD::// / 1::// C.1	<pre>ax.set_title('text', loc, pad)</pre>	
Title / subtitle of the plot	<pre>plt.suptitle('text', x, y, size, ha)</pre>	
Title of the $x / y$ axis	<pre>ax.set_xlabel('text') / ax.set_ylabel('text')</pre>	
Title of the size / color	<pre>ax.get_legend_handles_labels()</pre>	
Caption of the plot	<pre>ax.text('text', x, y, fontsize)</pre>	

This results in the following plot:



□ Double axes – A plot can have more than one axis with the plt.twinx() command. It is done as follows:

```
Python
ax2 = plt.twinx()
```

- ☐ Figure saving There are two main steps to save a plot:
  - Specifying the width and height of the plot when declaring the figure:

```
Python
f, ax = plt.subplots(1, figsize=(width, height))
```

• Saving the figure itself:

```
Python
f.savefig(fname)
```

SECTION 4

### Engineering productivity tips with Git, Bash and Vim

# 4.1 Working in groups with Git

## 4.1.1 Overview

- □ Overview Git is a version control system (VCS) that tracks changes of different files in a given repository. In particular, it is useful for:
  - · keeping track of file versions
  - working in parallel thanks to the concept of branches
  - · backing up files to a remote server

#### 4.1.2 Main commands

 $\square$  Getting started – The table below summarizes the commands used to start a new project, depending on whether or not the repository already exists:

Case	Action	Command	Illustration
No existing repository	Initialize repository from local folder	git init	<b>!</b> ⊋.
Repository already exists	Copy repository from remote to local	git clone git_address	path/to/address.git

□ File check-in — We can track modifications made in the repository, done by either modifying, adding or deleting a file, through the following steps:

Step	Command	Illustration
1. Add modified, new, or deleted file to staging area	git add file	
2. Save snapshot along with descriptive message	git commit -m 'description'	description

Remark 1: git add . will have all modified files to the staging area.

Remark 2: files that we do not want to track can be listed in the .gitignore file.

□ Sync with remote – The following commands enable changes to be synchronized between remote and local machines:

Action	Command	Illustration
Fetch most recent changes from remote branch	git pull name_of_branch	name_of_branch
Push latest local changes to remote branch	git push name_of_branch	name_of_branch

□ Parallel workstreams – In order to make changes that do not interfere with the current branch, we can create another branch name\_of\_branch as follows:

```
Bash
git checkout -b name_of_new_branch  # Create and checkout to that branch
```

Depending on whether we want to incorporate or discard the branch, we have the following commands:

Action	Command	Illustration
Merge with initial branch	git merge initial_branch	current_branch on name_of_branch
Remove branch	git branch -D name_of_branch	name_of_branch (

 $\hfill\Box$  Tracking status – We can check previous changes made to the repository with the following commands:

Action	Command	Illustration
Check status of modified file(s)	git status	Staged changes Unstaged changes Untracked files
View last commits	git logoneline	hf06f35 Change logic dey07e0 Rename folder 18887nd Update file
Compare changes made between two commits	<pre>git diff commit_1 commit_2</pre>	commit_1 commit_2  + added line - removed_line
View list of local branches	git branch	*current_branch other_branch_1 other_branch_2

□ Canceling changes – Canceling changes is done differently depending on the situation that we are in. The table below sums up the most common cases:

Case	Action	Command	Illustration
Unstaged	Revert file to last commit	git checkout file	file
Staged	Remove file from staging area	git reset HEAD file	file
Committed	Go back to a previous commit	git resethard prev_commit	prev_commit  HEAD HEAD

# 4.1.3 Project structure

 $\hfill \square$  Structure of folders – It is important to keep a consistent and logical structure of the project. One example is as follows:

```
Terminal

my_project/
analysis/
graph/
notebook/
data/
```

```
query/
  raw/
  processed/
modeling/
  method/
  tests
README.md
```

# 4.2 Working with Bash

□ Basic terminal commands – The table below sums up the most useful terminal commands:

Category	Action	Command
	Display list of files (including hidden ones)	ls (-a)
Exploration	Show current directory	pwd
	Show content of file	cat path_to_file
	Show statistics of file (lines/words/characters)	wc path_to_file
	Make new folder	mkdir folder_name
	Change directory to folder	cd path_to_folder
	Create new empty file	touch filename
File	Copy-paste file (folder) from origin to destination	scp (-R) origin destination
management	Move file/folder from origin to destination	mv origin destination
	Remove file (folder)	rm (-R) path
G .	Compress folder into file	tar -czvf comp_folder.tar.gz folder
Compression	Uncompress file	tar -xzvf comp_folder.tar.gz
	Display message	echo "message"
Miscellaneous	Overwrite / append file with output	<pre>output &gt; file.txt / output &gt;&gt; file.txt</pre>
	Execute command with elevated privileges	sudo command
	Connect to a remote machine	ssh remote_machine_address

 $\Box$  Chaining – It is a concept that improves readability by chaining operations with the pipe | operator. The most common examples are summed up in the table below:

Action	Command	
Count number of files in a folder	ls path_to_folder   wc -l	
Count number of lines in file	cat path_to_file   wc -l	
Show last n commands executed	history   tail -n	

□ Advanced search – The find command allows the search of specific files and manipulate them if necessary. The general structure of the command is as follows:

```
Bash find path_to_folder/. [conditions] [actions]
```

The possible conditions and actions are summarized in the table below:

Category	Action	Command
	Certain names, regex accepted	-name 'certain_name'
Conditions	Certain file types (d/f for directory/file)	-type certain_type
Conditions	Certain file sizes $(c/k/M/G \text{ for } B/kB/MB/GB)$	-size file_size
	Opposite of a given condition	-not [condition]
A	Delete selected files	-delete
Actions	Print selected files	-print

Remark: the flags above can be combined to make a multi-condition search.

 $\Box$  Changing permissions – The following command enables to change the permissions of a given file (or folder):

```
Bash chmod (-R) three_digits file
```

with three\_digits being a combination of three digits, where:

- the first digit is about the owner associated to the file
- the second digit is about the group associated to the file
- the third digit is anyone irrespective of their relation to the file

Each digit is one of (0, 4, 5, 6, 7), and has the following meaning:

Representation	Binary	Digit	Explanation
	000	0	No permission
r	100	4	Only read permission
r-x	101	5	Both read and execution permissions
rw-	110	6	Both read and write permissions
rwx	111	7	Read, write and execution permissions

For instance, giving read, write, execution permissions to everyone for a given\_file is done by running the following command:

```
Bash chmod 777 given_file
```

Remark: in order to change ownership of a file to a given user and group, we use the command chown user:group file.

 $\square$  Terminal shortcuts – The table below summarizes the main shortcuts when working with the terminal:

Action	Command
Search previous commands	Ctrl + r
Go to beginning / end of line	Ctrl + a / Ctrl + e
Remove everything after the cursor	Ctrl + k
Clear line	Ctrl + u
Clear terminal window	Ctrl + l

## 4.3 Automating tasks

 $\hfill \Box$  Create aliases – Shortcuts can be added to the ~/.bash\_profile file by adding the following code:

```
Bash shortcut="command"
```

 $\square$  Bash scripts – Bash scripts are files whose file name ends with .sh and where the file itself is structured as follows:

```
Bash
#!/bin/bash
... [bash script] ...
```

□ Crontabs – By letting the day of the month vary between 1-31 and the day of the week vary between 0-6 (Sunday-Saturday), a crontab is of the following format:

```
Terminal

* * * * * * *

minute hour day month day

of month of week
```

□ tmux − Terminal multiplexing, often known as tmux, is a way of running tasks in the background and in parallel. The table below summarizes the main commands:

Category	Action	Command
	Open a new / last existing session	tmux / tmux attach
Session management	Leave current session	tmux detach
	List all open sessions	tmux ls
	Remove session_name	tmux kill-session -t session_name
Window monomonat	Open / close a window	Cmd + b + c / Cmd + b + x
Window management	Move to $n^{\text{th}}$ window	Ctrl + b + n

## 4.4 Mastering editors

□ Vim – Vim is a popular terminal editor enabling quick and easy file editing, which is particularly useful when connected to a server. The main commands to have in mind are summarized in the table below:

Category	Action	Command
	Go to beginning / end of line	0 / \$
	Go to first / last line / $i^{\mathrm{th}}$ line	gg / G / i G
File handling	Go to previous / next word	b / w
	Exit file with / without saving changes	:wq / :q!
Text editing	Copy line n line(s), where $n \in \mathbb{N}$	nyy
Text cutting	Insert n line(s) previously copied	р
Searching	Search for expression containing name_of_pattern	/name_of_pattern
Scarcining	Next / previous occurrence of name_of_pattern	n / N
Replacing	Replace old with new expressions with confirmation for each change	:%s/old/new/gc

 $\square$  Jupyter notebook – Editing code in an interactive way is easily done through Jupyter notebooks. The main commands to have in mind are summarized in the table below:

Category	Action	Command
	Transform selected cell to text / code	Click on cell $+ m / y$
Cell transformation	Delete selected cell	Click on cell + dd
	Add new cell below / above selected cell	Click on cell + b / a

SECTION A

# Conversion between R and Python: data manipulation

## A.1 Main concepts

 $\hfill {\fill}$  File management – The table below summarizes the useful commands to make sure the working directory is correctly set:

Category	R Command	Python Command
	setwd(path)	os.chdir(path)
Paths	getwd()	os.getcwd()
	<pre>file.path(path_1,, path_n)</pre>	os.path.join(path_1,, path_n)
	<pre>list.files(    path, include.dirs = TRUE )</pre>	os.listdir(path)
Files	<pre>file_test('-f', path)</pre>	os.path.isfile(path)
	<pre>file_test('-d', path)</pre>	os.path.isdir(path)
	read.csv(path_to_csv_file)	pd.read_csv(path_to_csv_file)
	<pre>write.csv(df, path_to_csv_file)</pre>	df.to_csv(path_to_csv_file)

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

Category	R Command	Python Command
	df %>% select(col_list)	df[col_list]
Look at data	<pre>df %&gt;% head(n) / df %&gt;% tail(n)</pre>	<pre>df.head(n) / df.tail(n)</pre>
	df %>% summary()	df.describe()
Data types	df %>% str()	<pre>df.dtypes / df.info()</pre>
	df %>% NROW() / df %>% NCOL()	df.shape

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

R Data type	Python Data type	Description
character		String-related data
factor	object	String-related data that can be put in bucket, or ordered
numeric	float64	Numerical data
int	int64	Numeric data that are integer
POSIXct	datetime64	Timestamps

## A.2 Data preprocessing

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

```
df %>%
  filter(some_col some_operation some_value_or_list_or_col)
```

where some\_operation is one of the following:

Category	R Command	Python Command
	== / !=	== / !=
Basic	<, <=, >=, >	<, <=, >=, >
	& /	& /
	is.na()	pd.isnull()
Advanced	%in% (val_1,, val_n)	.isin([val_1,, val_n])
	%like% 'val'	.str.contains('val')

□ Mathematical operations – The table below sums up the main mathematical operations that can be performed on columns:

Operation	R Command	Python Command
$\sqrt{x}$	sqrt(x)	np.sqrt(x)
$\lfloor x \rfloor$	floor(x)	np.floor(x)
$\lceil x \rceil$	ceiling(x)	np.ceil(x)

### A.3 Data frame transformation

 $\hfill\Box$  Common transformations – The common data frame transformations are summarized in the table below:

Category	R Command	Python Command
Concatenation	rbind(df_1,, df_n)	pd.concat([df_1,, df_n], axis=0)
	cbind(df_1,, df_n)	<pre>pd.concat([df_1,, df_n], axis=1)</pre>
Dimension change	<pre>spread(df, key, value)</pre>	<pre>pd.pivot_table(   df, values='some_values',   index='some_index',   columns='some_column',   aggfunc=np.sum )</pre>
	<pre>gather(df, key, value)</pre>	<pre>pd.melt(    df, id_vars='variable',    value_vars='other_variable' )</pre>

- SECTION B

Conversion between R and Python: data visualization

# B.1 General structure

☐ Basic plots – The main basic plots are summarized in the table below:

Type	R Command	Python Command
Scatter plot	<pre>geom_point(    x, y, params )</pre>	<pre>sns.scatterplot(    x, y, params )</pre>
Line plot	<pre>geom_line(    x, y, params )</pre>	<pre>sns.lineplot(    x, y, params )</pre>
Bar chart	<pre>geom_bar(    x, y, params )</pre>	<pre>sns.barplot(    x, y, params )</pre>
Box plot	<pre>geom_boxplot(    x, y, params )</pre>	<pre>sns.boxplot(    x, y, params )</pre>
Heatmap	<pre>geom_tile(    x, y, params )</pre>	<pre>sns.heatmap(   x, y, params )</pre>

where the meaning of parameters are summarized in the table below:

Command	Description	Use case
color / hue	Color of a line / point / border	'red'
fill	Color of an area	'red'
size	Size of a line / point	4
linetype	Shape of a line	'dashed'
alpha	Transparency, between 0 and 1	0.3

# B.2 Advanced features

☐ Additional elements – We can add objects on the plot with the following commands:

Type	R Command	Python Command
Line	<pre>geom_vline(    xintercept, linetype )</pre>	<pre>ax.axvline(   x, ymin, ymax, color,   linewidth, linestyle )</pre>
	<pre>geom_hline(   yintercept, linetype )</pre>	<pre>ax.axhline(   y, xmin, xmax, color,   linewidth, linestyle )</pre>
Rectangle	<pre>geom_rect(    xmin, xmax, ymin, ymax )</pre>	ax.axvspan( xmin, xmax, ymin, ymax )
Text	<pre>geom_text(    x, y, label, hjust, vjust )</pre>	ax.text(    x, y, s, color )