MSc - Data Mining

Topic 01: Module Overview

Part 06 : Top X pandas commands

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Outline

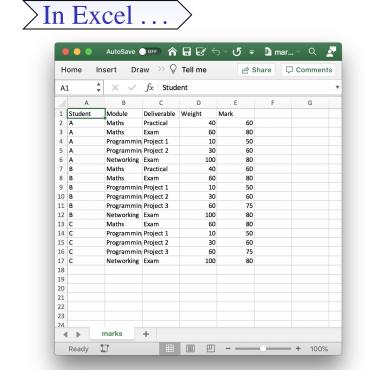
- Reading data formats
- Computing descriptive statistics
- Processing data by filtering and grouping

Part I

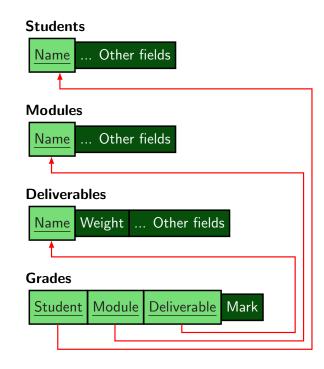
Introduction

Minimal Dataset

To better understand the various pandas operations we are going use a tiny* dataset based on (fictional) student results. (marks.csv)







$\rangle \dots$ like to know $\dots \rangle$

- Student performance —
 weighted mark on each
 module, missing deliverables
 etc.
- Module performance number of attempts and average mark.
- Deliverable performance number of attempts and average mark, predictor of overall module grade, etc.

^{*}Dataset is small enough that you can verify operation results by hand.

Terminology

0 1 2	A A	Module Maths	Deliverable	Weight	Monk
1		Maths		_	Mark
-	Α		Practical	40	60
2		Maths	Exam	60	80
	Α	Programming	Project 1	10	50
3	Α	Programming	Project 2	30	60
4	Α	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	С	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	С	Programming	Project 2	30	60
14	С	Programming	Project 3	60	75
15	С	Networking	Lab Work	100	80

- A DataFrame is a table of data values.
 - df = pd.read_csv("marks.csv")
- A Series is a list of data values typically columns in a dataframe. We can access an individual column using
 - df.Deliverable
 df["Deliverable"]
 df.iloc[:,2]
 (dict notation)
 (numpy, index notation)
- The index is a special column whose values can be used to access rows rather using row number.
 - The default index is equal to the row number.

Part II

Input and Output

Setup

Minimal

We begin every data mining project with importing the three core data science packages:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
plt.style.use('seaborn-darkgrid')
numpy — fast array operations
pandas — data manipulation
matplotlib — visualisation
```

• We give modules nicknames (np, pd, ...) to simplify their later use, and we access properties/functions of a package using the dot notation (np.max, pd.DataFrame, ...).

>Extra >

```
import seaborn as sns
import statsmodels.api as sm
pd.set_option('display.max_columns', 500)
pd.set_option('display.width', 1000)
seaborn — statistical visualisation
statsmodels — statistical data exploration
pandas options to show all columns for wider datasets
```

Reading data from a CSV file

Pandas supports a huge variety of input/output formats so best approach is to focus on what is needed to process the given data and verify input. Our marks dataset is in CSV format so we start with

```
import numpy as np
 import pandas as pd
 import matplotlib.pyplot as plt
 plt.style.use('seaborn-darkgrid')
                                                                                    (16, 5)
and input using
                                                                                     Student
                                                                                              Module Deliverable Weight Mark
                                                                                              Maths
                                                                                                    Practical
                                                                                              Maths
 df = pd.read_csv('marks.csv', sep=',')
                                                                                         A Programming
                                                                                                    Project 1
 print(df.shape)
                                                                                            Networking
                                                                                                    Lab Work
 df.head()
                                                   (16, 5)
```

Always verify input by checking dataset dimensions and looking at some rows!!!

Datatypes

Pandas data types:

- object used for text or mixed numeric and non-numeric values.
- int64 integer values,

• Does not support missing values, so an int column containing at least one missing value will

automatically be converted to float.

• float64 — floating point numbers.

• bool — True/False values

datetime64 — date and time values

• category — Finite (typically small) list of text values

Student object
Module object
Deliverable object
Weight int64
Mark int64
dtype: object

df.dtypes

Regularly verifying datatypes is vital[†]:

- Operations differ based on datatype, eg, '+' concatenate strings but adds numerical values.
- Datatype can change based on results, eg, int converts to float due to missing values.

[†]Google "Detecting Excel's gene auto-conversions."

Datatypes — Converting

We will deal with modifying and creating new columns later, but while we are on datatypes, we will look at changing datatype . . .

Using the Series function astype

```
df["Weight"] = df["Weight"].astype('float')
df["Weight"].dtype

dtype('float64')
```

- New datatype is required argument 'int', 'float', 'str', 'object', 'category', etc.
- Simple, but fragile if data conversion is possible.

```
or using pandas function to_numeric
```

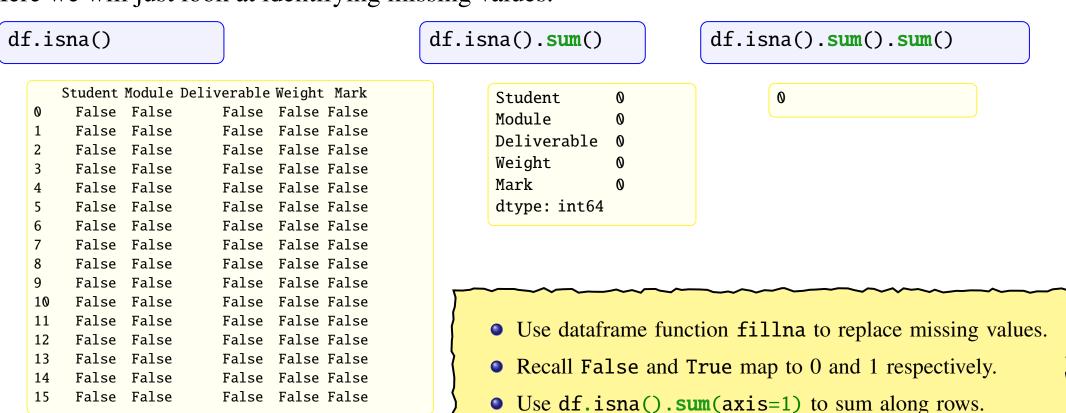
```
df["Weight"] = pd.to_numeric(df["Weight"])
df["Weight"].dtype

dtype('float64')
```

- More powerful, can specify what to do in cases where the conversion fails etc
- Have functions to_numeric, to_datetime, and to_timedelta.

Missing Values

Identifying and dealing with missing values is critical step in data preparation. What should you do? delete rows containing missing values? or impute then? Here we will just look at identifying missing values.



Output

Saving dataframe to CSV is straightforward (I rarely include the (default) index when saving datasets).

```
df.to_csv('marks_2.csv', index=False)
```

- CSV has become the default file format in Data Mining application especially for 'informal' datasets.
 - ✓ human readable, easy to generate / parse (if correct).
 - X Can be highly redundant, slow to input/output.
 - X No meta information.
- Other formats are better for speed and resulting file size and for saving meta data not supported by CSV (such as columns datatypes, category information, etc).

towards data science: The Best Format to Save Pandas Data

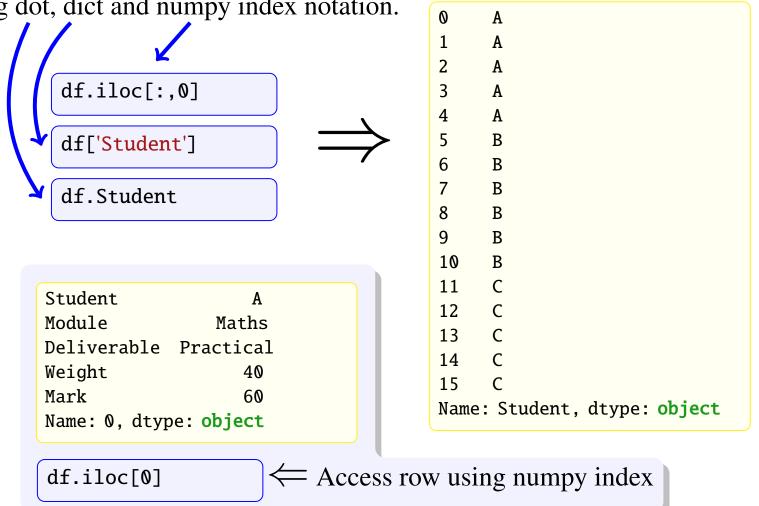
Part III

Filtering

Selecting individual rows/columns results in a series

Columns can accessed using dot, dict and numpy index notation.

df.head(1000)							
	Student	Module	Deliverable	Weight	Mark		
0	Α	Maths	Practical	40	60		
1	Α	Maths	Exam	60	80		
2	Α	Programming	Project 1	10	50		
3	Α	Programming	Project 2	30	60		
4	Α	Networking	Lab Work	100	80		
5	В	Maths	Practical	40	60		
6	В	Maths	Exam	60	80		
7	В	Programming	Project 1	10	50		
8	В	Programming	Project 2	30	60		
9	В	Programming	Project 3	60	75		
10	В	Networking	Project	100	80		
11	С	Maths	Exam	60	80		
12	С	Programming	Project 1	10	50		
13	С	Programming	Project 2	30	60		
14	С	Programming	Project 3	60	75		
15	С	Networking	Lab Work	100	80		



Head and Tail

Commands head and tail return the first and last n rows (default n = 5) of a dataframe/series.

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	С	Programming	Project 3	60	75
15	С	Networking	Lab Work	100	80

df.head()

Student	Module	Deliverable	Weight	Marl
<u>0</u> A	Maths	Practical	40	60
1 A	Maths	Exam	60	80
2 A	Programming	Project 1	10	50
3 A	Programming	Project 2	30	60
4 A	Networking	Lab Work	100	80

df.tail()

Studen	t Module	Deliverable	Weight	Mark
11 C	Maths	Exam	60	80
12 C	Programming	Project 1	10	50
13 C	Programming	Project 2	30	60
14 C	Programming	Project 3	60	75
15 C	Networking	Lab Work	100	80

df.head(1)

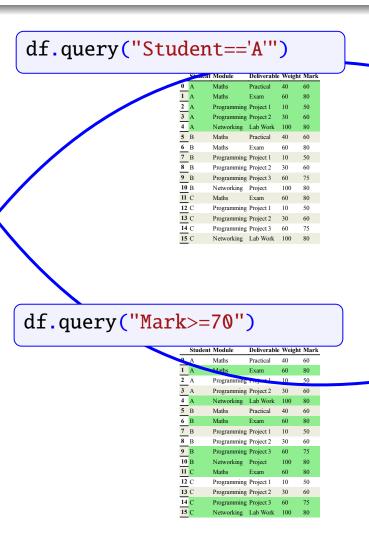
Student	Module	Deliverable	Weight	Mark
<u>0</u> A	Maths	Practical	40	60

df.tail()

	Student	Module	Deliverable	Weight	Mark
15	C	Networking	Lab Work	100	80

Query — on a single-column criteria

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	С	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80



Student	t Module	Deliverable	Weight	Mark
<u>0</u> A	Maths	Practical	40	60
1 A	Maths	Exam	60	80
2 A	Programming	Project 1	10	50
3 A	Programming	Project 2	30	60
4 A	Networking	Lab Work	100	80

	Student	Module	Deliverable	Weight	Mark
1	A	Maths	Exam	60	80
4	A	Networking	Lab Work	100	80
6	В	Maths	Exam	60	80
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
14	С	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

Query — on multiple columns (using python logical operators)

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	C	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

	Student	Module	Deliverable	Weight	Mar
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	C	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	С	Maths	Exam	60	80
12	C	Programming	Project 1	10	50
13	С	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

df.query("Mark<70 and Module=='Maths'")</pre>

Student	Module	Deliverable	Weight	Marl
<u>0</u> A	Maths	Practical	40	60
5 B	Maths	Practical	40	60

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60

df.query("Mark<70 or Module=='Maths'")</pre>

Query — on multiple columns (using pandas logical operators)

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	C	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	С	Networking	Lab Work	100	80

	Student	Module	Deliverable	Weight	Mar
0	A	Maths	Practical	40	60
1_	A	Maths	Exam	60	80
2_	A	Programming	Project 1	10	50
3_	A	Programming	Project 2	30	60
4_	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80
	Student	Module	Deliverable	Weight	Marl

_	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	С	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	С	Programming	Project 2	30	60
14	С	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

df.query("(Mark<70) & (Module=='Maths')")</pre>

Student	Module	Deliverable	Weight	Marl
<u>0</u> A	Maths	Practical	40	60
5 B	Maths	Practical	40	60

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
11	C	Maths	Exam	60	80
12	C	Programming	Project 1	10	50
13	С	Programming	Project 2	30	60

df.query("(Mark<70) | (Module=='Maths')")</pre>

Filtering using 10c

Note the square (not round) brackets — think of loc as array indexing not a function call.

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	C	Programming	Project 3	60	75
15	С	Networking	Lab Work	100	80

df.loc[ROW_SELECTION, COL_SELECTION] where row and columns selection can be

- Single values: row number or column name
- An integer list for rows or list of column names
- A boolean list for logical indexing of rows
- A colon to indicate every row/column

7 B	Programming	Project 1	10	50
8 B	Programming	Project 2	30	60
9 B	Programming	Project 3	60	75
10 B	Networking	Project	100	80
11 C	Maths	Exam	60	80

1	2 C	Programming	Project 1	10	50
1	13 C	Programming	Project 2	30	60
1	14 C	Programming	Project 3	60	75
1	15 C	Networking	Lab Work	100	80

Student Mark

0	A	60
1	A	80
5	В	60
6	В	80
11	C	80

More complicated example

I prefer to define row selection criteria, and the column list and order, separately to the loc statement.

_	Module Maths	Deliverable	Weight	Mark		0 A 1 A 2 A	Maths Maths Programming	Practical Exam Project 1	40 60 10	60 80 50						
A N			Weight	Mark		2 A			10	50						
A N			Weight	Mark		-										
_	Maths					3 A	Programming	Project 2	30	60			nt Module	Deliverabl	e Weight	Mark
_	viauis	Drootical	40	60		4 A	Networking	Lab Work	100	80		<u>0</u> A	Maths	Practical	40	60
_A N		Practical	40			6 B	Maths Maths	Practical Exam	60	60 80		1 A	Maths	Exam	60	80
	Maths	Exam	60	80		7 B	Programming	.,	10	50		2 A	Programming	Project 1	10	50
A P	Programming	Project 1	10	50		8 B 9 B	Programming Programming	.,	30 60	75		3 A	Programming	Project 2	30	60
_						10 B	Networking	Project	100	80		4 A	Networking	Lab Work	100	80
_A P	Programming	Project 2	30	60		11 C 12 C	Maths Programming	Exam Project 1	60 10	50		5 B	Maths	Practical	40	60
A N	Networking	Lab Work	100	80		13 C	Programming	,	30	60		6 B	Maths	Exam	60	80
_	Maths	Practical	40	60		14 C 15 C	Programming Networking	-3	60	75 80		7 B	Programming	Project 1	10	50
_		Flactical				13 (Networking	Lab work	100	80		8 B	Programming		30	60
_B N	Maths	Exam	60	80								9 B	Programming		60	75
B P	Programming	Project 1	10	50								10 B	Networking		100	80
- В Р	Programming	Project 2	30	60								11 C	Maths	Exam	60	80
_			60	75								12 C	Programming	g Project 1	10	50
В Р	Programming	Project 3	60	13								13 C	Programming	Project 2	30	60
				-) & (df.Mod lent', 'Mark		e=='N	ſatl	1S	'))	(df.Mark<70) & (d:	f.Mod	ıle!=	='Ma	th

Sampling

The sample function selects a random subset of the dataframe rows.

- Either specify the number of rows (as an integer) or fraction of the data (as a float).
- Can set the seed using random_state parameter for reproducible samples.

	Student	Module	Deliverable	Weight	Mark
0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
3	A	Programming	Project 2	30	60
4	A	Networking	Lab Work	100	80
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
8	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
10	В	Networking	Project	100	80
11	C	Maths	Exam	60	80
12	С	Programming	Project 1	10	50
13	C	Programming	Project 2	30	60
14	С	Programming	Project 3	60	75
15	C	Networking	Lab Work	100	80

	Student	Module	Deliverable	Weight	Mark
11	С	Maths	Exam	60	80
3	A	Programming	Project 2	30	60
8	В	Programming	Project 2	30	60

	Student	Module	Deliverable	Weight	Mark
8	В	Programming	Project 2	30	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50

	Student	Module	Deliverable	Weight	Mark
10	В	Networking	Project	100	80
6	В	Maths	Exam	60	80
2	A	Programming	Project 1	10	50

Part IV

Sorting

Sorting

A pandas dataframe has two sorting operations:

- sort_index() orders rows based on current index.
- sort_values(COLUMNS) orders rows based on single column or list of columns.

Two important modifications:

- By default, the sort order is in ascending. Set parameter ascending=False to reverse this.
- By default, a new dataframe is returned with desired sort order, set parameter inplace=True to update current dataframe instead (then no output is generated).

	Student	Module	Deliverable	Weight	Mark		Student	Module	Deliverable	Weight	
0	A	Maths	Practical	40	60		1 A	Maths	Exam	60	
1	A	Maths	Exam	60	80		6 B	Maths	Exam	60	
2	A	Programming	Project 1	10	50		11 C	Maths	Exam	60	
3	A	Programming	Project 2	df.	sor	t_values(['Module','Deliverable'])	A	Maths	Practical	40	
4	A	Networking	Lab Work			, , , , , , , , , , , , , , , , , , ,	БВ	Maths	Practical	40	
5	В	Maths	Practical	40	60		4 A	Networking	Lab Work	100	
6	В	Maths	Exam	60	80		15 C	Networking	Lab Work	100	
7	В	Programming	Project 1	10	50		10 B	Networking	Project	100	
8	В	Programming	Project 2	30	60		2 A	Programming	Project 1	10	
9	В	Programming	Project 3	60	75		7 B	Programming	Project 1	10	
10	В	Networking	Project	100	80		12 C	Programming	Project 1	10	
11	С	Maths	Exam	60	80		3 A	Programming	Project 2	30	
12		Programming	Project 1	10	50		8 B	Programming	Project 2	30	
13	С	Programming	Project 2	30	60		13 C	Programming	Project 2	30	
14		Programming	Project 3	60	75		9 B	Programming	Project 3	60	
15	С	Networking	Lab Work	100	80		14 C	Programming	Project 3	60	

Part V

Defining New Columns

Defining new columns — row-wise operation

We want to compute the weighted mark for each module for each student. Two steps:

- Create column, W_Mark, to store the weighted mark for each deliverable. This is a row by row calculation only need data in current row to compute the result.
- Create column, M_Mark, to store the module mark for each student. This is a group calculation need all rows for that student and module to compute the result.

Student	Module	Deliverable	Weigh	t Mark						St	udent	Module	Deliverable	Weight	Mark	. W_Mark
1 A	Maths	Exam	60	80						1 A		Maths	Exam	60	80	48
6 B	Maths	Exam	60	80						6 B		Maths	Exam	60	80	48
11 C	Maths	Exam	60	80						11 C		Maths	Exam	60	80	48
0 A	Maths	Practical	40	60						0 A		Maths	Practical	40	60	24
5 B	Maths	Practical	40	CEITA	M	-1-!7	-1 C	Tara di sala d	+ . dC Mal- /	/ 100		Maths	Practical	40	60	24
4 A	Networking	Lab Work	10 C	II [W_	_Mar	CK']	= ai.	weign	t * df.Mark //	TOO		Networking	Lab Work	100	80	80
15 C	Networking	Lab Work	100	80						15 C		Networking	Lab Work	100	80	80
10 B	Networking	Project	100	80					>	10 B		Networking	Project	100	80	80
2 A	Programming	Project 1	10	50						2 A		Programming	Project 1	10	50	5
7 B	Programming	Project 1	10	50	,	\sim			~~~~	7 B		Programming	Project 1	10	50	5
12 C	Programming	Project 1	10	50	(Ne	eed to	use dic	t notation (not	12 C		Programming	Project 1	10	50	5
3 A	Programming	Project 2	30	60	((3 A		Programming	Project 2	30	60	18
8 B	Programming	Project 2	30	60	,				hen defining a	8 B		Programming	Project 2	30	60	18
13 C	Programming	Project 2	30	60	>) ne	w colu	ımn.		13 C		Programming	Project 2	30	60	18
9 B	Programming	Project 3	60	75	L		~~	\sim		9 B		Programming	Project 3	60	75	45
<u>14</u> C	Programming	Project 3	60	75						14 C		Programming	Project 3	60	75	45

Defining new columns — group aggregate result

- Create column, W_Mark, to store the weighted mark for each deliverable. This is a row by row calculation only need data in current row to compute the result.
- Create column, M_Mark, to store the module mark for each student. This is a group calculation need all rows for that student and module to compute the result.

columns to group on output cols aggregate df.groupby(['Student','Module'])['W_Mark'].sum()

	Student	Module	Deliverable	Weight	Mark	W_Mark
1	A	Maths	Exam	60	80	48
6	В	Maths	Exam	60	80	48
11	С	Maths	Exam	60	80	48
0	A	Maths	Practical	40	60	24
5	В	Maths	Practical	40	60	24
4	A	Networking	Lab Work	100	80	80
15	С	Networking	Lab Work	100	80	80
10	В	Networking	Project	100	80	80
2	A	Programming	Project 1	10	50	5
7	В	Programming	Project 1	10	50	5
12	C	Programming	Project 1	10	50	5
3	A	Programming	Project 2	30	60	18
8	В	Programming	Project 2	30	60	18
13	С	Programming	Project 2	30	60	18
9	В	Programming	Project 3	60	75	45
14	C	Programming	Project 3	60	75	45

Result has multi-level index, need to use reset_index to revert to default index

		w_Mark
Student	Module	
	Maths	72
A	Networking	80
	Programming	23
	Maths	72
В	Networking	80
	Programming	68
	Maths	48
C	Networking	80
	Programming	68

W Moule

Defining new columns — group aggregate result

```
columns to group on output cols aggregate df.groupby(['Student','Module'])[['W_Mark']].sum().reset_index()
```

	Student	Module	Deliverable	Weight	Mark	W_Mark
1	A	Maths	Exam	60	80	48
6	В	Maths	Exam	60	80	48
11	C	Maths	Exam	60	80	48
0	A	Maths	Practical	40	60	24
5	В	Maths	Practical	40	60	24
4	A	Networking	Lab Work	100	80	80
15	C	Networking	Lab Work	100	80	80
10	В	Networking	Project	100	80	80
2	A	Programming	Project 1	10	50	5
7	В	Programming	Project 1	10	50	5
12	C	Programming	Project 1	10	50	5
3	A	Programming	Project 2	30	60	18
8	В	Programming	Project 2	30	60	18
13	C	Programming	Project 2	30	60	18
9	В	Programming	Project 3	60	75	45
14	C	Programming	Project 3	60	75	45

This is the required result and we can save this to a new dataframe. However, we often want to put this into to out original dataframe as an extra column. Only problem we have different rows so can't just assign to a new column — need to use transform function.

	Student	Module	$\mathbf{W}_{\underline{}}$	_Mark
0	A	Maths	72	
1	A	Networking	80	
2	A	Programming	23	
3	В	Maths	72	
4	В	Networking	80	
5	В	Programming	68	
6	C	Maths	48	
7	C	Networking	80	
	C	Programming	68	

```
columns to group on output cols aggregate df['M_Mark'] = df.groupby(['Student','Module'])[['W_Mark']].transform(sum)
```

Stude	ent Module	Deliverable	e Weigh	t Mark	W_Mark						Stude	nt Module	Deliverable	Weight	t Marl	W_Mark	M_Mark
1 A	Maths	Exam	60	80	48					1	A	Maths	Exam	60	80	48	72
6 B	Maths	Exam	60	80	48					6	В	Maths	Exam	60	80	48	72
11 C	Maths	Exam	60	80	48		\			11	C	Maths	Exam	60	80	48	48
0 A	Maths	Practical	40	60	24					0	A	Maths	Practical	40	60	24	72
5 B	Maths	Practical	40	60	24		•			5	В	Maths	Practical	40	60	24	72
4 A	Networking	Lab Work	100	80						4	A	Networking	Lab Work	100	80	80	80
15 C	Networking	Lab Work	100	80			nt Module	W_Mark		15	C	Networking	Lab Work	100	80-	80	80
10 B	Networking	Project	100	80	4	0 A 1 A	Maths Networking	72		10	В	Networking	Project	100	80	80	1 80
2 A	Programming	Project 1	10	50		$\frac{1}{2}$ A	Programming			2	K	Programming	Project 1	10	50	5	23
7 B	Programming	Project 1	10	50		$\frac{2}{3}$ B	Maths	72	- II.	7	В	Programming	Project 1	10	50	5	68
12 C	Programming	Project 1	10	50		4 B	Networking			12	C	Programming	Project 1	10	50	5	68
3 A	Programming	Project 2	30	60		5 B	Programming			3	A	Programming	Project 2	30	60	18	23
8 B	Programming	Project 2	30	60		6 C	Maths	48		8	В	Programming	Project 2	30	60	18	68
13 C	Programming	-	30	60		7 C	Networking	80		13	C	Programming	Project 2	30	60	18	68
9 B	Programming	Project 3	60	75	4	8 C	Programming	68		9	В	Programming	Project 3	60	75	45	68
14 C	Programming		60	75	۷. پ					14	C	Programming	Project 3	60	75	45	68

The transform broadcasts the result for each group over every row in that group.

Part VI

Review Exercises

Review Exercises

Generate the following reports:

- Number of deliverables by each student.
- List and rank deliverables by grade.
- Top 2 deliverables (by grade).
- Top 2 module (by average grade).
- Top 2 modules (by minimum grade).
- Modules (by minimum grade).

Harder exercises (new functions)

• List which students missed which deliverables.

(value_counts, or groupby and count)

(sort_values, rank)

(pivot, melt)