(MSc) Data Mining

Topic 01: Module Overview

Part 13: Top X pandas commands

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Spring Semester, 2025

Outline

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

Part I

Introduction

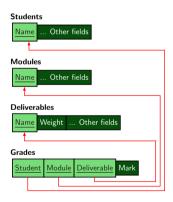
Minimal Dataset

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





... or database schema ...



...like to know ...

- 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

df.	df.head(1000)							
Г	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	C	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 (dot notation
 - df["Deliverable"] (dict notation
 - df.iloc[:,2] (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.

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8	В	Programming	Project 2	30	60
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10	В	Networking	Project	100	80
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			1		

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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)
```

seaborn — statistical visualisation statsmodels — statistical data exploration pandas options to show all columns for wider datasets

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Extra

```
import seaborn as sns
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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
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and input using

```
df = pd.read_csv('Marks.csv', sep=',')
print(df.shape)
df.head()
```

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```
import numby as no
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 import matplotlib.pyplot as plt
 plt.style.use('seaborn-darkgrid')
                                                                           (16.5)
and input using
                                                                                          Practica
                                                                                                    60
                                                                                     Maths
                                                                                           Exam
 df = pd.read_csv('Marks.csv', sep=',')
                                                                                          Project 1
 print(df.shape)
 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."

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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.

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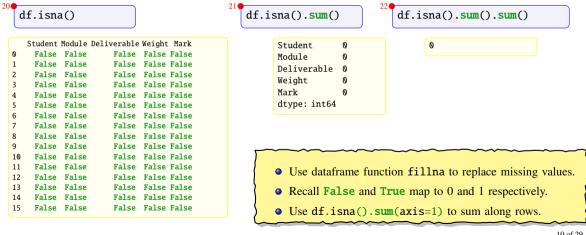
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Missing Values

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).

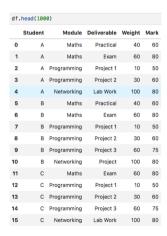
towards data science: The Best Format to Save Pandas Data

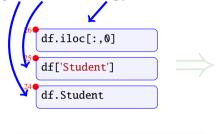
Part III

Filtering

Selecting individual rows/columns results in a series

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



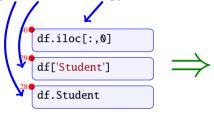


Student	A
Module	Maths
Deliverable	Practical
Weight	40
Mark	60
Name: 0, dty	

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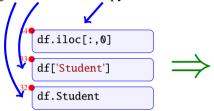
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```
Α
      Α
      Α
      Α
      Α
      В
      В
      В
      В
      В
10
      В
11
12
13
14
15
Name: Student, dtype: object
```

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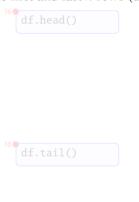
Student A
Module Maths
Deliverable Practical
Weight 40
Mark 60
Name: 0, dtype: object

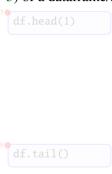
```
Α
      Α
      Α
      Α
      В
      В
      В
      R
10
      R
11
12
13
14
15
Name: Student, dtype: object
```

Head and Tail

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

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1	A	Maths	Exam	60	80
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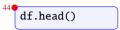
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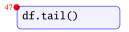
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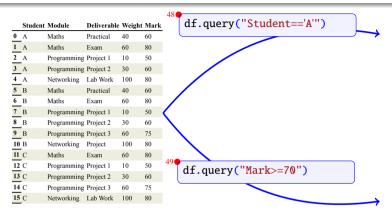


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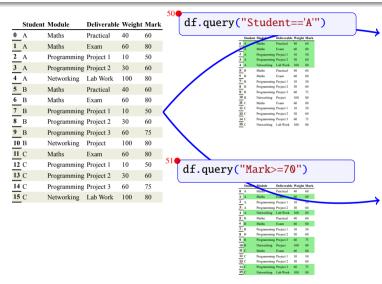


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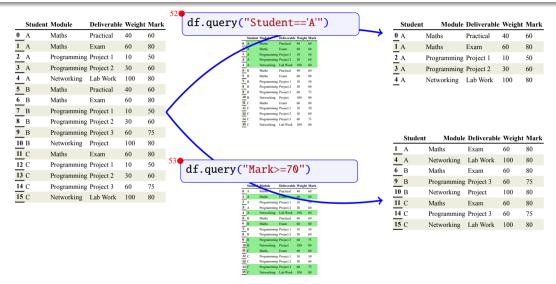
Query — on a single-column criteria



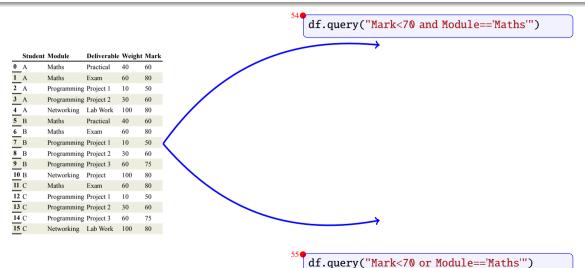
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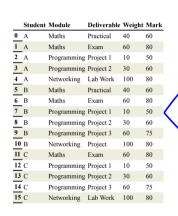
Query — on a single-column criteria



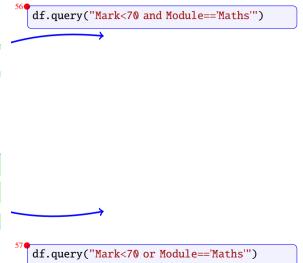
Query — on multiple columns (using python ogical operators)



Query — on multiple columns (using python ogical operators)







Query — on multiple columns (using python logical operators)

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0	A	Maths	Practical	40	60
1	A	Maths	Exam	60	80
2	A	Programming	Project 1	10	50
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5	В	Maths	Practical	40	60
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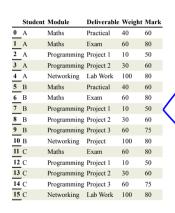
58						
df.query("Mar	k<	<70 a	nd Mo	dule==	'Math	s'")
\rightarrow	5	Student	Module	Deliverable	Weight	Mark
	0 /	A	Maths	Practical	40	60
	5 1	В	Maths	Practical	40	60
		Student	Mode	ale Deliverab	le Weigh	Mark
	0	A	Maths	Practical	40	60
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	7	В	Programmi	ng Project 1	10	50
	8	В	Programmi	ng Project 2	30	60
\longrightarrow	11	C	Maths	Exam	60	80
	12	C	Programmi	ng Project 1	10	50

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

Programming Project 2

13 C

Query — on multiple columns (using pandas ogical operators)





\rightarrow						
		Student	Module I	Deliverable	Weight	Mark
	0	A	Maths P	ractical	40	60
	5	В	Maths P	Practical	40	60
		Student	Modu	le Deliverab	le Weigh	t Mark
	0		Maths	Practical	40	60
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	2	Α	Programmir	ng Project 1	10	50
	2	Λ.	Programmir	ng Project 2	30	60
	3	. ^	1 10B14111111			
	5	В	Maths	Practical	40	60
	2 3 5 6	B B		Practical Exam	40 60	60 80
	7	В	Maths	Exam		
	7	В	Maths Maths	Exam ng Project 1	60	80
	7	В	Maths Maths Programmir	Exam ng Project 1	60 10	80 50
<u></u>	7	B B B C C	Maths Maths Programmir Programmir	Exam ng Project 1 ng Project 2 Exam	60 10 30	80 50 60

Filtering using 10c

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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

```
df.loc[df.Module=="Maths", ["Student", "Mark"]]
```

Filtering using 10c

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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

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

df.loc[df.Module=="Maths", ["Student", "Mark"]]

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	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.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



	Student	Mark
0	A	60
1	A	80
5	В	60
6	В	80
11	С	80

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More complicated example

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

	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

```
criteria = ((df.Mark<50) & (df.Module=='Maths')) | ((df.Mark<70) & (df.Module!='Maths'))
columns = ['Module', 'Student', 'Mark']
df.loc[criteria, columns]</pre>
```

More complicated example

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

	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

6 B Maths Exam 60 38 7 B Programming Project 1 10 50 8 B Programming Project 2 30 66 9 B Programming Project 3 60 75 10 B Networking Project 1 100 86 11 C Maths Exam 60 80	
A Programming Project 10 54 A Programming Project 20 64 A Networking Project 20 64 5 B Matho Project 40 65 6 B Matho Exam 60 85 7 B Programming Project 10 65 8 Programming Project 20 64 8 Programming Project 50 65 9 B Programming Project 60 71 10 B Networking Project 100 84 11 C Matho Exam 60 85 12 C Matho Exam 60 85 13 Programming Project 100 84 14 T T T T T T 15 T T T T T T 16 T T T T T T 17 T T T T T 18 T T T T T 19 T T T T T 10 T T T T 10 T T T T 10 T	
2 A Programme Project 1 to 5 5 4 A Networking Lub Work 100 6 6 1 Mallo Project 4 0 0 6 1 Mallo Project 4 0 0 0 0 0 0 0 0 0	0
5 II Maths Practical 40 66 6 B Maths Exam 60 86 7 B Programming Project 1 10 58 8 B Programming Project 2 30 66 9 B Programming Project 3 60 73 10 B Networking Project 1 100 38 11 C Maths Exam 60 80	0
5 II Maths Practical 40 66 6 B Maths Exam 60 86 7 B Programming Project 1 10 58 8 B Programming Project 2 30 66 9 B Programming Project 3 60 73 10 B Networking Project 1 100 38 11 C Maths Exam 60 80	0
5 II Maths Practical 40 66 6 B Maths Exam 60 86 7 B Programming Project 1 10 58 8 B Programming Project 2 30 66 9 B Programming Project 3 60 73 10 B Networking Project 1 100 38 11 C Maths Exam 60 80	0
6 B Maths Exam 60 38 7 B Programming Project 1 10 50 8 B Programming Project 2 30 66 9 B Programming Project 3 60 75 10 B Networking Project 1 100 86 11 C Maths Exam 60 80	0
8 B Programming Project 2 30 66 9 B Programming Project 3 60 75 10 B Networking Project 100 80 11 C Maths Exam 60 80	0
9 B Programming Project 3 60 75 10 B Networking Project 100 80 11 C Maths Exam 60 80	0
10 B Networking Project 100 80 11 C Maths Exam 60 80	0
II C Maths Exam 60 80	8
	0
	0
12 C Programming Project 1 10 50	0
13 C Programming Project 2 30 66	0
14 C Programming Project 3 60 75	5
15 C Networking Lab Work 100 80	0

```
criteria = ((df.Mark<50) & (df.Module=='Maths')) | ((df.Mark<70) & (df.Module!='Maths'))
columns = ['Module', 'Student', 'Mark']
df.loc[criteria, columns]</pre>
```

More complicated example

Deliverable Weight Mark

Student Module

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

						4 A	Networking I	ab Work 10	0 81	0		0 A	Maths	Practical	40	60
_	Maths	Practical	40	60		4 A 5 B	Maths P	Practical 40	64	0		1 A	Maths	Exam	60	80
1 A	Maths	Exam	60	80		6 B 7 B 8 B	Programming P	Project I 16	51	0		2 A	Programming	Project 1	10	50
2 A	Programming	Project 1	10	50			Programming P	Project 2 30 Project 3 66	2	0 S		3 A	Programming		30	60
_	Programming	-		60		9 B 10 B	Networking P	Project 10	0 81	0	•	4 A	Networking	Lab Work	100	80
_						11 C 12 C 13 C	Maths I Programming P	Ixam 66 Project I 10	51	0		5 B	Maths	Practical	40	60
4_A	Networking	Lab Work	100	80	4	13 C	Programming P	roject 2 30	64	0		6 B	Maths	Exam	60	80
5 B	Maths	Practical	40	60		14 C 15 C	Programming P Networking L	hojeet 3 66 ab Work 16	0 81	S 0		7 B	Programming	Project 1	10	50
6 B	Maths	Exam	60	80		_						8 B	Programming	Project 2	30	60
_												9 B	Programming	Project 3	60	75
7 B	Programming	Project 1	10	50								10 B	Networking	Project	100	80
В	Programming	Project 2	30	60								11 C	Maths	Exam	60	80
В В	Programming	Project 2	60	75								12 C	Programming	Project 1	10	50
	1 logramming	1 Toject 5	00	15								13 C	Programming	Project 2	30	60
			-		(df.Mod		=='Ma	aths	s'))	((df.Mark<70)	& (df	. Modu	le!=	'Mat	hs
f.loc[criteri	ia, col	Lumn	s]												

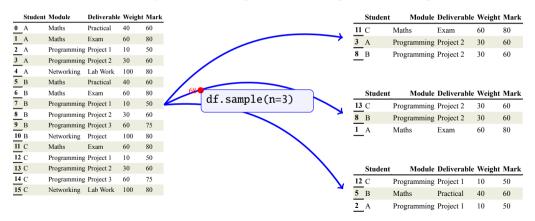
Deliverable Weight Mark

Student Module

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.



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
0	A	Maths	Practical	40	60
	A	Maths	Exam	60	80
2	A A	Programming	Project 69	10	50
3	A	Programming	Project 2	df.	sor
	A	Networking	Lab Work		
5	В	Maths	Practical	40	60
6	В	Maths	Exam	60	80
7	В	Programming	Project 1	10	50
	В	Programming	Project 2	30	60
9	В	Programming	Project 3	60	75
	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

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).

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

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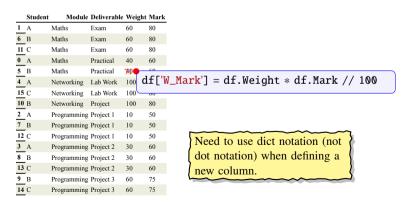
Part V

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.



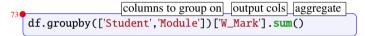
Defining new columns — row-wise operation

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- 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	Weig	ht Mark	<u>.</u>								Studen	Module	Deliverable	Weight	Mark	W_Mar
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	710			-								Maths	Practical	40	60	24
4 A	Networking	Lab Work	100	df['I	W_Mai	ark	[] =	= df	.Wei	ght	* df.Marl	c //	100	letworking	Lab Work	100	80	80
_	Networking	Lab Work	100	00	_	_							15 C	Networking	Lab Work	100	80	80
10 B	Networking	Project	100	80								7	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									7 B	Programming	Project 1	10	50	5
	Programming	Project 1	10	50		N	Jaar	d to	1100 (lict r	notation (no)t	12 C	Programming	Project 1	10	50	5
3 A	Programming	Project 2	30	60	S)							3 A	Programming	Project 2	30	60	18
8 B	Programming	Project 2	30	60	l	d	ot 1	notai	tion)	whe	n defining	a)	8 B	Programming	Project 2	30	60	18
13 C	Programming	Project 2	30	60	1	l n	ew	colu	umn.			Y	13 C	Programming	Project 2	30	60	18
9 B	Programming	Project 3	60	75	,	1		~					9 B	Programming	Project 3	60	75	45
14 C	Programming	Project 3	60	75									14 C	Programming	Project 3	60	75	45

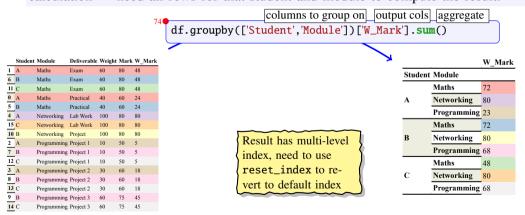
- Create column, W_Mark, to store the weighted mark for each deliverable. This is a row by row calculation only need data in current row to compute the result.
- Create column, M_Mark, to store the module mark for each student. This is a group calculation need all rows for that student and module to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



	Student	Module	Deliverable	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	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		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	С	Programming	Project 3	60	75	45

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

- 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().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	С	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

```
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	С	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	С	Programming	Project 3	60	75	45	

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

_	Student	Module	W_Mark
0	A	Maths	72
1	A	Networking	80
2	A	Programming	23
3	В	Maths	72
	В	Networking	80
5	В	Programming	68
6	C	Maths	48
7	C	Networking	80
8	C	Programming	68

III

columns to group on output cols aggregate df['M_Mark'] = df.groupby(['Student','Module'])[['W_Mark']].transform(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	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	С	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	С	Programming	Project 3	60	75	45

_	_		
		N	
		4	
	Student	Module	W_Mark
	A	Maths	72
1	A	Networking	80
2	A A B	Programming	23
3	В	Maths	72
4	В	Networking	80
5	B C C	Programming	68
6	C	Maths	48
7	C	Networking	80
8	C	Programming	68

		Student	Module	Deliverable	Weight	Mark	W_Mark	M_Mark
	1	A	Maths	Exam	60	80	48	72
	6	В	Maths	Exam	60	80	48	72
	11	С	Maths	Exam	60	80	48	48
	0	A	Maths	Practical	40	60	24	72
	5	В	Maths	Practical	40	60	24	72
	4	A	Networking	Lab Work	100	80	80	80
	15	C	Networking	Lab Work	100	80	80	80
	10	В	Networking	Project	100	80	80	80
	2	K	Programming	Project 1	10	50	5	23
'	7	В	Programming	Project 1	10	50	5	68
	12	C	Programming	Project 1	10	50	5	68
	3	A	Programming	Project 2	30	60	18	23
	8	В	Programming	Project 2	30	60	18	68
	13	C	Programming	Project 2	30	60	18	68
	9	В	Programming	Project 3	60	75	45	68
	14	С	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.
- **1** Top 2 deliverables (by grade).
- Top 2 module (by average grade).
- **5** 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)