

## p7\_and\_p8

September 15, 2023

p7

```
[ ]: import numpy as np
      dtypes = [('emp_id', int), ('last_name', 'U20'), ('first_name', 'U20'),
      ↪ ('gender', 'U20'), ('Title', 'U20')]

      employee_table = np.array([], dtype=dtypes)
      emp1 = (1000, "Trbati", "Yolanda", "F", "Programmer")
      emp2 = (1001, "Kleinn", "Joel", "M", "Programmer")
      emp3 = (1002, "Ginsburg", "Laura", "F", "President")
      emp4 = (1003, "Cox", "Jennifer", "F", "Programmer")
      emp5 = (1005, "Ziada", "Mauri", "M", "Product Designer")
      emp6 = (1006, "Keyser", "Cara", "F", "Account Executive")
      emp7 = (1010, "Smith", "Roxie", "M", "Programmer")
      emp8 = (1011, "Nelson", "Robert", "M", "Programmer")
      emp9 = (1012, "Sachsen", "Lars", "M", "Support Technician")
      emp10 = (1013, "Shannon", "Don", "M", "Product Designer")

      employee_table = np.append(employee_table, np.array([emp1, emp2,
      ↪ emp3, emp4, emp5, emp6, emp7, emp8, emp9, emp10], dtype=dtypes))
      print(employee_table)
```

```
[(1000, 'Trbati', 'Yolanda', 'F', 'Programmer')
 (1001, 'Kleinn', 'Joel', 'M', 'Programmer')
 (1002, 'Ginsburg', 'Laura', 'F', 'President')
 (1003, 'Cox', 'Jennifer', 'F', 'Programmer')
 (1005, 'Ziada', 'Mauri', 'M', 'Product Designer')
 (1006, 'Keyser', 'Cara', 'F', 'Account Executive')
 (1010, 'Smith', 'Roxie', 'M', 'Programmer')
 (1011, 'Nelson', 'Robert', 'M', 'Programmer')
 (1012, 'Sachsen', 'Lars', 'M', 'Support Technician')
 (1013, 'Shannon', 'Don', 'M', 'Product Designer')]
```

1. How many Male employees are in a company?

```
[ ]: male_emp = employee_table[employee_table['gender'] == 'M']
      nm= len(male_emp)
      print("Number of Male employees in the Company are :",nm)
```

Number of Male employees in the Company are : 6

2.Display the details of employees whose Last\_Name starts with S.

```
[ ]: s_emp = employee_table[np.char.startswith(employee_table['last_name'], 'S')]  
print(s_emp)
```

```
[(1010, 'Smith', 'Roxie', 'M', 'Programmer')  
 (1012, 'Sachsen', 'Lars', 'M', 'Support Technician')  
 (1013, 'Shannon', 'Don', 'M', 'Product Designer')]
```

3.Sort the Female Employee details in descending order based on First\_Name

```
[ ]: sf_emp = np.sort(employee_table, order='first_name')[::-1]  
print(sf_emp)
```

```
[(1000, 'Trbati', 'Yolanda', 'F', 'Programmer')  
 (1010, 'Smith', 'Roxie', 'M', 'Programmer')  
 (1011, 'Nelson', 'Robert', 'M', 'Programmer')  
 (1005, 'Ziada', 'Mauri', 'M', 'Product Designer')  
 (1002, 'Ginsburg', 'Laura', 'F', 'President')  
 (1012, 'Sachsen', 'Lars', 'M', 'Support Technician')  
 (1001, 'Kleinn', 'Joel', 'M', 'Programmer')  
 (1003, 'Cox', 'Jennifer', 'F', 'Programmer')  
 (1013, 'Shannon', 'Don', 'M', 'Product Designer')  
 (1006, 'Keyser', 'Cara', 'F', 'Account Executive')]
```

4.Extract 1D array and reshape it into 2D array.

```
[ ]: empid_1d = employee_table['emp_id']  
empid_2d = empid_1d.reshape(-1, 1)  
print("1-D array was ",empid_1d)  
print("2-D array is ",empid_2d)
```

1-D array was [1000 1001 1002 1003 1005 1006 1010 1011 1012 1013]

2-D array was [[1000]

[1001]

[1002]

[1003]

[1005]

[1006]

[1010]

[1011]

[1012]

[1013]]

5.Extract the below matrix using Boolean and Fancy indexing.

```
[ ]: start_emp_id = 1002  
end_emp_id = 1012  
mask = (employee_table['emp_id'] >= start_emp_id) & (employee_table['emp_id']  
↪ <= end_emp_id)
```

```
columns_to_extract = ['emp_id', 'last_name', 'Title']
filtered_data = employee_table[mask][columns_to_extract]
print(filtered_data)
```

```
((1002, 'Ginsburg', 'President') (1003, 'Cox', 'Programmer')
(1005, 'Ziada', 'Product Designer') (1006, 'Keyser', 'Account Executive')
(1010, 'Smith', 'Programmer') (1011, 'Nelson', 'Programmer')
(1012, 'Sachsen', 'Support Technician'))
```

p8

1. Import the domain dataset that you identified with missing values and perform the following.

For each output, give the interpretation with respect to the imported dataset.

```
[ ]: import numpy as np
import pandas as pd
d1=pd.read_csv("C:/Users/Hp/Downloads/archive (2)/student-mat.csv")
```

2.Read the csv file and create and understand the data frame using describe(), shape, info().

```
[ ]: print(d1)
print("DataFrame Description:")
print(d1.describe())
print("\nDataFrame Shape:")
print(d1.shape)
print("\nDataFrame Info:")
print(d1.info())
```

	school	sex	age	address	famsize	Pstatus	Medu	Fedu	Mjob	Fjob	
0	GP	F	18	U	GT3	A	4	4	at_home	teacher	\
1	GP	F	17	U	GT3	T	1	1	at_home	other	
2	GP	F	15	U	LE3	T	1	1	at_home	other	
3	GP	F	15	U	GT3	T	4	2	health	services	
4	GP	F	16	U	GT3	T	3	3	other	other	
..	...	..	...	...	...	...	...	...	...	...	
390	MS	M	20	U	LE3	A	2	2	services	services	
391	MS	M	17	U	LE3	T	3	1	services	services	
392	MS	M	21	R	GT3	T	1	1	other	other	
393	MS	M	18	R	LE3	T	3	2	services	other	
394	MS	M	19	U	LE3	T	1	1	other	at_home	

	...	famrel	freetime	goout	Dalc	Walc	health	absences	G1	G2	G3
0	...	4	3	4	1	1	3	6	5	6	6
1	...	5	3	3	1	1	3	4	5	5	6
2	...	4	3	2	2	3	3	10	7	8	10
3	...	3	2	2	1	1	5	2	15	14	15
4	...	4	3	2	1	2	5	4	6	10	10
..	...	...	...	...	...	...	...	...	...	...	...
390	...	5	5	4	4	5	4	11	9	9	9

391	...	2	4	5	3	4	2	3	14	16	16
392	...	5	5	3	3	3	3	3	10	8	7
393	...	4	4	1	3	4	5	0	11	12	10
394	...	3	2	3	3	3	5	5	8	9	9

[395 rows x 33 columns]

DataFrame Description:

	age	Medu	Fedu	traveltime	studytime	failures	
count	395.000000	395.000000	395.000000	395.000000	395.000000	395.000000	\
mean	16.696203	2.749367	2.521519	1.448101	2.035443	0.334177	
std	1.276043	1.094735	1.088201	0.697505	0.839240	0.743651	
min	15.000000	0.000000	0.000000	1.000000	1.000000	0.000000	
25%	16.000000	2.000000	2.000000	1.000000	1.000000	0.000000	
50%	17.000000	3.000000	2.000000	1.000000	2.000000	0.000000	
75%	18.000000	4.000000	3.000000	2.000000	2.000000	0.000000	
max	22.000000	4.000000	4.000000	4.000000	4.000000	3.000000	

	famrel	freetime	goout	Dalc	Walc	health	
count	395.000000	395.000000	395.000000	395.000000	395.000000	395.000000	\
mean	3.944304	3.235443	3.108861	1.481013	2.291139	3.554430	
std	0.896659	0.998862	1.113278	0.890741	1.287897	1.390303	
min	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	
25%	4.000000	3.000000	2.000000	1.000000	1.000000	3.000000	
50%	4.000000	3.000000	3.000000	1.000000	2.000000	4.000000	
75%	5.000000	4.000000	4.000000	2.000000	3.000000	5.000000	
max	5.000000	5.000000	5.000000	5.000000	5.000000	5.000000	

	absences	G1	G2	G3
count	395.000000	395.000000	395.000000	395.000000
mean	5.708861	10.908861	10.713924	10.415190
std	8.003096	3.319195	3.761505	4.581443
min	0.000000	3.000000	0.000000	0.000000
25%	0.000000	8.000000	9.000000	8.000000
50%	4.000000	11.000000	11.000000	11.000000
75%	8.000000	13.000000	13.000000	14.000000
max	75.000000	19.000000	19.000000	20.000000

DataFrame Shape:

(395, 33)

DataFrame Info:

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 395 entries, 0 to 394

Data columns (total 33 columns):

#	Column	Non-Null Count	Dtype
0	school	395 non-null	object
1	sex	395 non-null	object

```

2   age          395 non-null   int64
3   address      395 non-null   object
4   famsize      395 non-null   object
5   Pstatus      395 non-null   object
6   Medu         395 non-null   int64
7   Fedu         395 non-null   int64
8   Mjob         395 non-null   object
9   Fjob         395 non-null   object
10  reason       395 non-null   object
11  guardian     395 non-null   object
12  traveltime   395 non-null   int64
13  studytime    395 non-null   int64
14  failures     395 non-null   int64
15  schoolsup    395 non-null   object
16  famsup       395 non-null   object
17  paid         395 non-null   object
18  activities   395 non-null   object
19  nursery      395 non-null   object
20  higher       395 non-null   object
21  internet     395 non-null   object
22  romantic     395 non-null   object
23  famrel       395 non-null   int64
24  freetime     395 non-null   int64
25  goout        395 non-null   int64
26  Dalc         395 non-null   int64
27  Walc         395 non-null   int64
28  health       395 non-null   int64
29  absences     395 non-null   int64
30  G1           395 non-null   int64
31  G2           395 non-null   int64
32  G3           395 non-null   int64
dtypes: int64(16), object(17)
memory usage: 102.0+ KB
None

```

3. Find if any missing values (null values) are in the data, handle all the rows with missing data in four different ways (delete, replace, fill, bill), and print the data frame.

```

[ ]: df_deleted = d1.dropna()
     df_replaced = d1.fillna(-1)
     df_filled_forward = d1.ffill()
     df_filled_backward = d1.bfill()

```

4. Filter based on any column using groupby().

```

[ ]: grouped_data = d1.groupby('Mjob')['age'].mean()
     print(grouped_data)

```

Mjob

```

at_home      16.966102
health       16.352941
other        16.751773
services     16.679612
teacher      16.517241
Name: age, dtype: float64

```

5. Select 20 samples randomly and Create a data frame with Hierarchical Index

```

[ ]: random_samples = d1.sample(n=20)
      hierarchical_df = random_samples.set_index(['school', 'sex'])
      print("\nDataFrame with Hierarchical Index:")
      print(hierarchical_df)

```

DataFrame with Hierarchical Index:

		age	address	famsize	Pstatus	Medu	Fedu	Mjob	Fjob	
school	sex									
GP	F	15	U	LE3	A	4	3	other	other	\
	F	16	U	LE3	T	2	4	other	health	
MS	F	19	R	GT3	T	2	3	services	other	
GP	M	16	U	LE3	T	4	3	health	other	
	F	16	R	GT3	T	3	3	services	other	
	M	16	U	LE3	T	2	1	other	other	
MS	F	20	U	GT3	T	4	2	health	other	
GP	M	17	R	LE3	T	1	1	other	services	
	M	17	U	LE3	T	4	4	other	teacher	
	F	15	U	GT3	T	1	2	at_home	other	
MS	M	17	U	LE3	T	3	1	services	services	
GP	M	17	R	GT3	T	1	2	at_home	other	
	F	15	U	GT3	A	3	3	other	health	
	F	16	U	LE3	T	4	4	health	health	
	M	16	U	GT3	T	3	2	at_home	other	
	F	18	R	LE3	T	1	1	at_home	other	
	M	15	U	GT3	T	4	4	teacher	health	
	F	15	U	LE3	A	3	4	other	other	
	F	17	U	LE3	T	0	2	at_home	at_home	
MS	F	18	R	GT3	T	1	1	other	other	

		reason	guardian	...	famrel	freetime	goout	Dalc	Walc	
school	sex			...						
GP	F	course	mother	...	5	2	2	1	1	\
	F	course	father	...	4	2	2	1	2	
MS	F	course	mother	...	5	4	2	1	2	
GP	M	home	father	...	3	1	3	1	3	
	F	reputation	father	...	4	1	2	1	1	
	M	course	mother	...	4	2	3	1	2	
MS	F	course	other	...	5	4	3	1	1	
GP	M	course	mother	...	5	3	5	1	5	

	M	home	father	...	4	1	1	2	2
	F	course	mother	...	4	3	2	1	1
MS	M	course	mother	...	2	4	5	3	4
GP	M	home	mother	...	3	1	3	1	5
	F	reputation	father	...	4	3	3	1	1
	F	other	mother	...	5	4	5	1	1
	M	reputation	mother	...	5	3	3	1	3
	F	reputation	mother	...	5	2	2	1	1
	M	reputation	mother	...	3	2	2	1	1
	F	home	mother	...	5	3	2	1	1
	F	home	father	...	3	3	3	2	3
MS	F	home	mother	...	4	3	2	1	2

		health	absences	G1	G2	G3
school	sex					
GP	F	5	8	8	8	6
	F	5	2	13	13	13
MS	F	5	0	7	5	0
GP	M	5	4	8	10	10
	F	2	0	7	10	10
	M	5	0	15	15	15
MS	F	3	4	15	14	15
GP	M	5	0	5	8	7
	M	5	0	11	11	10
	F	5	2	10	11	11
MS	M	2	3	14	16	16
GP	M	3	4	8	9	10
	F	4	10	10	11	11
	F	4	4	14	15	16
	M	2	10	11	9	10
	F	3	1	12	12	12
	M	5	4	14	15	15
	F	1	0	7	10	11
	F	2	0	16	15	15
MS	F	4	2	8	8	10

[20 rows x 31 columns]