

```

1 """
2 1, Upload the data file, view the data file in Colab
3 2, Read and Assign it to a variable called users and use the 'user_id' as index
4 3, See the first 25 entries
5 4, See the last 10 entries
6 5, What is the number of observations in the dataset?
7 6, What is the number of columns in the dataset?
8 7, Print the name of all the columns.
9 8, How is the dataset indexed?
10 9, What is the data type of each column?
11 10, Print only the occupation column
12 11, How many different occupations are in this dataset?
13 12, What is the most frequent occupation?
14 13, Summarize the DataFrame.
15 14, Summarize all the columns.
16 15, Summarize only the occupation column?
17 16, What is the mean age of users?
18 17, What is the age with least occurrence?
19 18, Discover what is the mean age per occupation
20 19, Discover the Male ratio per occupation and sort it from the most to the least
21 20, For each occupation, calculate the minimum and maximum ages
22 21, For each combination of occupation and gender, calculate the mean age
23 22, For each occupation present the percentage of women and men
24 """

```

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1, Upload the data file, view the data file in Colab

done

```

1 import numpy as np
2 import pandas as pd
3 occ= pd.read_csv("Occupation.csv")
4 occ

```

	user_id	age	gender	occupation	zip_code
0				1 24 M technician 85711	
1				2 53 F other 94043	
2				3 23 M writer 32067	
3				4 24 M technician 43537	
4				5 33 F other 15213	
...				...	
938				939 26 F student 33319	

2, Read and Assign it to a variable called users and use the 'user_id' as index

940 941|20|M|student|97229

```
1 users= pd.read_csv("Occupation.csv",sep='|')
2 users
```

	user_id	age	gender	occupation	zip_code
0	1	24	M	technician	85711
1	2	53	F	other	94043
2	3	23	M	writer	32067
3	4	24	M	technician	43537
4	5	33	F	other	15213
...
938	939	26	F	student	33319
939	940	32	M	administrator	02215
940	941	20	M	student	97229
941	942	48	F	librarian	78209
942	943	22	M	student	77841

943 rows × 5 columns

```
1 users.index=users['user_id']
```

```
1 del users['user_id']
```

```
1 users
```

	age	gender	occupation	zip_code
user_id				
1	24	M	technician	85711
2	53	F	other	94043
3	23	M	writer	32067
4	24	M	technician	43537
5	33	F	other	15213
...
939	26	F	student	33319
940	32	M	administrator	02215
941	20	M	student	97229
942	48	F	librarian	78200

1 users.info()


```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 943 entries, 1 to 943
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         943 non-null   int64
1   gender      943 non-null   object
2   occupation  943 non-null   object
3   zip_code    943 non-null   object
dtypes: int64(1), object(3)
memory usage: 36.8+ KB
```

1 users.describe()

	age
count	943.000000
mean	34.051962
std	12.192740
min	7.000000
25%	25.000000
50%	31.000000
75%	43.000000
max	73.000000

3, See the first 25 entries

```
1 users.head(25)
```

	age	gender	occupation	zip_code	
user_id					
1	24	M	technician	85711	
2	53	F	other	94043	
3	23	M	writer	32067	
4	24	M	technician	43537	
5	33	F	other	15213	
6	42	M	executive	98101	
7	57	M	administrator	91344	
8	36	M	administrator	05201	
9	29	M	student	01002	
10	53	M	lawyer	90703	
11	39	F	other	30329	
12	28	F	other	06405	
13	47	M	educator	29206	
14	45	M	scientist	55106	
15	49	F	educator	97301	
16	21	M	entertainment	10309	
17	30	M	programmer	06355	
18	35	F	other	37212	
19	40	M	librarian	02138	
20	42	F	homemaker	95660	
21	26	M	writer	30068	
22	25	M	writer	40206	
23	30	F	artist	48197	
24	21	F	artist	94533	
25	39	M	engineer	55107	

4, See the last 10 entries

```
1 users.tail(10)
```

	age	gender	occupation	zip_code	
user_id					
934	61	M	engineer	22902	
935	42	M	doctor	66221	
936	24	M	other	32789	
937	48	M	educator	98072	
938	38	F	technician	55038	
939	26	F	student	33319	
940	32	M	administrator	02215	
941	20	M	student	97229	
942	48	F	librarian	78209	

5, What is the number of observations in the dataset?

```
1 users.shape
```

```
(943, 4)
```

```
1 users.shape[0]
```

```
943
```

6, What is the number of columns in the dataset?

```
1 users.shape[1]
```

```
4
```

7, Print the name of all the columns.

```
1 users.columns
```

```
Index(['age', 'gender', 'occupation', 'zip_code'], dtype='object')
```

8, How is the dataset indexed?

```
1 users.index
```

```
Int64Index([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10,
             ...
            934, 935, 936, 937, 938, 939, 940, 941, 942, 943],
           dtype='int64', name='user_id', length=943)
```

9, What is the data type of each column?

```
1 users.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 943 entries, 1 to 943
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         943 non-null    int64
1   gender      943 non-null    object
2   occupation  943 non-null    object
3   zip_code    943 non-null    object
dtypes: int64(1), object(3)
memory usage: 36.8+ KB
```

10, Print only the occupation column

```
1 users.occupation
```

```
user_id
1      technician
2         other
3         writer
4      technician
5         other
...
939        student
940  administrator
941        student
942        librarian
943        student
Name: occupation, Length: 943, dtype: object
```

11, How many different occupations are in this dataset?

12, What is the most frequent occupation?

```
1 users.occupation.describe()
```

```
count      943
unique       21
top      student
freq       196
Name: occupation, dtype: object
```

```
1 users.occupation.value_counts().count()
```

```
21
```

```
1 users.occupation.value_counts().count()
```

21

```
1 users.occupation.unique()
```

```
array(['technician', 'other', 'writer', 'executive', 'administrator',
      'student', 'lawyer', 'educator', 'scientist', 'entertainment',
      'programmer', 'librarian', 'homemaker', 'artist', 'engineer',
      'marketing', 'none', 'healthcare', 'retired', 'salesman', 'doctor'],
      dtype=object)
```

```
1 users.occupation.nunique()
```

21

```
1 occfreq= users.groupby('occupation')['occupation'].count()
2 occfreq
```

```
occupation
administrator    79
artist            28
doctor            7
educator          95
engineer          67
entertainment     18
executive         32
healthcare        16
homemaker         7
lawyer            12
librarian         51
marketing         26
none              9
other            105
programmer        66
retired           14
salesman          12
scientist         31
student          196
technician        27
writer            45
Name: occupation, dtype: int64
```

```
1 occfreq.sort_values()
```

```
occupation
doctor            7
homemaker         7
none              9
salesman          12
lawyer            12
retired           14
healthcare        16
entertainment     18
marketing         26
technician        27
artist            28
```

```
scientist      31
executive     32
writer         45
librarian     51
programmer    66
engineer      67
administrator 79
educator      95
other        105
student      196
Name: occupation, dtype: int64
```

13, Summarize the DataFrame.

```
1 users.describe()
```

	age
count	943.000000
mean	34.051962
std	12.192740
min	7.000000
25%	25.000000
50%	31.000000
75%	43.000000
max	73.000000

14, Summarize all the columns.

```
1 users.describe(include='all')
```


	age	gender	occupation	zip_code	
count	943.000000	943	943	943	

15, Summarize only the occupation column?

```
1 users.occupation.describe()
```

```
count      943
unique      21
top        student
freq       196
Name: occupation, dtype: object
```

16, What is the mean age of users?

```
1 users.age.mean()
```

```
34.05196182396607
```

17, What is the age with least occurrence?

```
1 users.age.describe()
```

```
count      943.000000
mean       34.051962
std        12.192740
min         7.000000
25%        25.000000
50%        31.000000
75%        43.000000
max        73.000000
Name: age, dtype: float64
```

```
1 agegroup = users.groupby('age')['age'].count()
2 agegroup
```

```
age
7      1
10     1
11     1
13     5
14     3
..
66     1
68     2
69     2
70     3
73     1
Name: age, Length: 61, dtype: int64
```

```
1 agegroup[agegroup==agegroup.min()]
age
7      1
10     1
11     1
66     1
73     1
Name: age, dtype: int64
```

8, Discover what is the mean age per occupation

```
1 users.groupby('occupation')['age'].mean().sort_index()

occupation
administrator    38.746835
artist            31.392857
doctor           43.571429
educator          42.010526
engineer          36.388060
entertainment     29.222222
executive         38.718750
healthcare        41.562500
homemaker         32.571429
lawyer            36.750000
librarian         40.000000
marketing         37.615385
none              26.555556
other             34.523810
programmer        33.121212
retired           63.071429
salesman          35.666667
scientist         35.548387
student           22.081633
technician        33.148148
writer            36.311111
Name: age, dtype: float64
```

19, Discover the Male ratio per occupation and sort it from the most to the least

```
1 pd.get_dummies(users.gender)
```

	F	M
user_id		
1	0	1
2	1	0
3	0	1
4	0	1
5	1	0

```
1 usersc=users.copy()
2 usersc
```

	age	gender	occupation	zip_code
user_id				
1	24	M	technician	85711
2	53	F	other	94043
3	23	M	writer	32067
4	24	M	technician	43537
5	33	F	other	15213
...
939	26	F	student	33319
940	32	M	administrator	02215
941	20	M	student	97229
942	48	F	librarian	78209
943	22	M	student	77841

943 rows × 4 columns

```
1 usersc=pd.concat([usersc, pd.get_dummies(users.gender)], axis='columns')
2 usersc
```

	age	gender	occupation	zip_code	F	M	
user_id							
1	24	M	technician	85711	0	1	
2	53	F	other	94043	1	0	
3	23	M	writer	32067	0	1	
4	24	M	technician	43537	0	1	
5	33	F	other	15213	1	0	

```
1 mratio=usersc.groupby('occupation')['M'].sum()/users.groupby('occupation')['gender'].c
2 mratio
```

```
occupation
administrator    0.544304
artist            0.535714
doctor            1.000000
educator          0.726316
engineer          0.970149
entertainment     0.888889
executive         0.906250
healthcare        0.312500
homemaker         0.142857
lawyer            0.833333
librarian         0.431373
marketing         0.615385
none              0.555556
other             0.657143
programmer        0.909091
retired           0.928571
salesman          0.750000
scientist         0.903226
student           0.693878
technician        0.962963
writer            0.577778
dtype: float64
```

```
1 mratio.sort_values(ascending=False)
```

```
occupation
doctor      1.000000
engineer    0.970149
technician  0.962963
retired     0.928571
programmer  0.909091
executive   0.906250
scientist   0.903226
entertainment 0.888889
lawyer      0.833333
salesman    0.750000
educator    0.726316
student     0.693878
other       0.657143
marketing   0.615385
writer      0.577778
none        0.555556
```


```
administrator    0.544304
artist           0.535714
librarian        0.431373
healthcare       0.312500
homemaker        0.142857
dtype: float64
```

alternate


```
1
```

20, For each occupation, calculate the minimum and maximum ages

```
1 occMinMax=pd.concat([
2     usersc.groupby('occupation')['age'].min(),
3     usersc.groupby('occupation')['age'].max()],
4     axis='columns')
5 occMinMax
```

	age	age	
occupation			
administrator	21	70	

```
1 occMinMax.columns=[ 'Min_Age', 'Max_Age' ]
2 occMinMax
```

	Min_Age	Max_Age	
occupation			
administrator	21	70	
artist	19	48	
doctor	28	64	
educator	23	63	
engineer	22	70	
entertainment	15	50	
executive	22	69	
healthcare	22	62	
homemaker	20	50	
lawyer	21	53	
librarian	23	69	
marketing	24	55	
none	11	55	
other	13	64	
programmer	20	63	
retired	51	73	
salesman	18	66	
scientist	23	55	
student	7	42	
technician	21	55	
writer	18	60	

```
1 #alternate
2 users.groupby('occupation')['age'].agg(['min', 'max'])
```

	min	max
occupation		
administrator	21	70
artist	19	48
doctor	28	64
educator	23	63
engineer	22	70
entertainment	15	50
executive	22	69
healthcare	22	62
homemaker	20	50
lawyer	21	53
librarian	23	69
marketing	24	55
none	11	55
other	13	64
programmer	20	63
retired	51	73
salesman	18	66
scientist	23	55
student	7	42
technician	21	55

21, For each combination of occupation and gender, calculate the mean age

```
1 round(usersc.groupby(['occupation','gender'])['age'].mean())
```

occupation	gender	
administrator	F	41.0
	M	37.0
artist	F	30.0
	M	32.0
doctor	M	44.0
educator	F	39.0
	M	43.0
engineer	F	30.0
	M	37.0
entertainment	F	31.0
	M	29.0
executive	F	44.0
	M	38.0
healthcare	F	40.0

	M	45.0
homemaker	F	34.0
	M	23.0
lawyer	F	40.0
	M	36.0
librarian	F	40.0
	M	40.0
marketing	F	37.0
	M	38.0
none	F	36.0
	M	19.0
other	F	35.0
	M	34.0
programmer	F	32.0
	M	33.0
retired	F	70.0
	M	63.0
salesman	F	27.0
	M	39.0
scientist	F	28.0
	M	36.0
student	F	21.0
	M	23.0
technician	F	38.0
	M	33.0
writer	F	38.0
	M	35.0

Name: age, dtype: float64

22, For each occupation present the percentage of women and men

```

1 occGen=pd.concat([
2     usersc.groupby('occupation')['M'].sum(),
3     usersc.groupby('occupation')['F'].sum()],
4     axis='columns')
5 occGen


```



	M	F
occupation		
administrator	43	36
artist	15	13
doctor	7	0
educator	69	26
engineer	65	2
entertainment	16	2
executive	29	3
healthcare	5	11
homemaker	1	6
lawyer	10	2
librarian	22	29



```
1 occGen[ 'Men%' ]=round(occGen[ 'M' ]/(occGen[ 'M' ]+occGen[ 'F' ]),2)*100
2 occGen[ 'WoMen%' ]=round(occGen[ 'F' ]/(occGen[ 'M' ]+occGen[ 'F' ]),2)*100
3 occGen
```





	M	F	Men%	WoMen%	
occupation					
administrator	43	36	54.0	46.0	
artist	15	13	54.0	46.0	
doctor	7	0	100.0	0.0	
educator	69	26	73.0	27.0	
engineer	65	2	97.0	3.0	
entertainment	16	2	89.0	11.0	
executive	29	3	91.0	9.0	
healthcare	5	11	31.0	69.0	
homemaker	1	6	14.0	86.0	
lawyer	10	2	83.0	17.0	
librarian	22	29	43.0	57.0	
marketing	16	10	62.0	38.0	
none	5	4	56.0	44.0	
other	69	36	66.0	34.0	
programmer	60	6	91.0	9.0	
retired	13	1	93.0	7.0	
salesman	9	3	75.0	25.0	