

```
In [1]: #uploading in jupyter notebook
import pandas as pd
import numpy as np
```

```
In [2]: pwd
```

```
Out[2]: 'C:\\Users\\DELL'
```

```
In [24]: df = pd.read_csv('NationalNames.csv.crdownload')
```

```
In [25]: df.head()
```

```
Out[25]:
```

	Id	Name	Year	Gender	Count
0	1	Mary	1880	F	7065.0
1	2	Anna	1880	F	2604.0
2	3	Emma	1880	F	2003.0
3	4	Elizabeth	1880	F	1939.0
4	5	Minnie	1880	F	1746.0

```
In [26]: #Show the first and last 10 rows
df.info
```

```
Out[26]: <bound method DataFrame.info of
Count
```

	Id	Name	Year	Gender
0	1	Mary	1880	F
1	2	Anna	1880	F
2	3	Emma	1880	F
3	4	Elizabeth	1880	F
4	5	Minnie	1880	F
...	...	...	...	...
932403	932404	Silpa	1983	F
932404	932405	Simmone	1983	F
932405	932406	Sinead	1983	F
932406	932407	Sioban	1983	F
932407	932408	Snow	1983	NaN

```
[932408 rows x 5 columns]>
```

```
In [27]: #there more male or female names in the dataset
df['Gender'].value_counts()
```

```
Out[27]: F    553010
M     379397
Name: Gender, dtype: int64
```

```
In [28]: len(df[(df['Name'] == 'Yes') & (df['Gender'] == 'Female')])
```

```
Out[28]: 0
```

```
In [29]: len(df['Gender'].value_counts())/len(df)*100
```

```
Out[29]: 0.0002144983741023243
```

```
In [12]: df.columns
```

```
Out[12]: Index(['Id', 'Name', 'Year', 'Gender', 'Count'], dtype='object')
```

```
In [30]: #the total counts of each Name  
df.groupby('Name').sum()
```

```
Out[30]:
```

	Id	Year	Count
<b>Name</b>			
<b>Aage</b>	127770	1915	7.0
<b>Aagot</b>	123994	1915	5.0
<b>Aaisha</b>	892274	1981	6.0
<b>Aakash</b>	1764335	3960	10.0
<b>Aaliyah</b>	6865111	15836	257.0
...	...	...	...
<b>Zygmund</b>	3202338	32652	168.0
<b>Zygmunt</b>	7375988	52092	363.0
<b>Zylpha</b>	1510688	26728	84.0
<b>Zylphia</b>	893923	9600	29.0
<b>Zyndall</b>	1155292	3919	14.0

45259 rows × 3 columns

```
In [14]: df = pd.read_csv('titanic_data.csv')
```

```
In [33]: import pandas as pd
```

```
In [51]: #Create a dataframe named bank_client_df as shown below using dictionary  
  
data = {  
    'ClientID': [1, 2, 3, 4, 5],  
    'Name': ['parnav', 'iti', 'prince', 'kevin', 'kasula'],  
    'net worth': [28, 35, 22, 40, 32],  
    'years': [5, 4, 5, 2, 6]  
}
```

```
In [52]: bank_client_df = pd.DataFrame(data)
```

In [53]: `print(bank_client_df)`

	ClientID	Name	net worth	years
0	1	parnav	28	5
1	2	iti	35	4
2	3	prince	22	5
3	4	kevin	40	2
4	5	kasula	32	6

In [54]: *#Define a function that increases all clients networkth (stocks) by a fixed*  
 apply it on Net Worth column  
*function that increases all clients networkth (stocks) by a fixed value of*  
 apply it on Net Worth column  
`def increase_net_worth(net_worth):`  
 `return net_worth * 1.10`

In [55]: `bank_client_df['net worth'] = bank_client_df['net worth'].apply(increase_net_worth)`

In [56]: `print(bank_client_df)`

	ClientID	Name	net worth	years
0	1	parnav	30.8	5
1	2	iti	38.5	4
2	3	prince	24.2	5
3	4	kevin	44.0	2
4	5	kasula	35.2	6

In [57]: *#total Networkth*  
`total_net_worth = bank_client_df['net worth'].sum()`  
`print("Total net worth: $", total_net_worth)`

Total net worth: \$ 172.7

In [58]: *#Filter rows where Years with bank is greater than or equal to 5 years.*

`filtered_df = bank_client_df[bank_client_df['years'] >= 5]`  
`print(filtered_df)`

	ClientID	Name	net worth	years
0	1	parnav	30.8	5
2	3	prince	24.2	5
4	5	kasula	35.2	6

In [59]: *#Rename Yash into Rahul*

```
bank_client_df['Name'] = bank_client_df['Name'].replace('kasula', 'kala papita')
print(bank_client_df)
```

	ClientID	Name	net worth	years
0	1	parnav	30.8	5
1	2	iti	38.5	4
2	3	prince	24.2	5
3	4	kevin	44.0	2
4	5	kala papita	35.2	6

In [71]: *#Read titanic\_data csv in jupyter notebook*

```
df = pd.read_csv('titanic_data.csv')
```

In [72]: `import pandas as pd`

In [73]: `df.head()`

Out[73]:

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male
--	----------	--------	-----	-----	-------	-------	------	----------	-------	-----	------------

0	0	3	male	22.0	1	0	7.2500	S	Third	man	True
1	1	1	female	38.0	1	0	71.2833	C	First	woman	False
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True

In [75]: `df.describe()`

Out[75]:

	survived	pclass	age	sibsp	parch	fare
count	889.000000	889.000000	713.000000	889.000000	889.000000	889.000000
mean	0.384702	2.307087	29.698696	0.523060	0.382452	32.259059
std	0.486799	0.836367	14.536691	1.103729	0.806761	49.735870
min	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	0.000000	2.000000	20.000000	0.000000	0.000000	7.925000
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

In [76]: *#total number of passengers from each class who survived?*

```
passengers_df = pd.DataFrame(data)
```

In [79]: `df.groupby('survived').sum()`

Out[79]:

	pclass	age	sibsp	parch	fare	adult_male	alone
survived							
0	1384	12955.50	303	181	12127.0741	447	373
1	667	8219.67	162	159	16551.2294	88	163

In [80]: *#the average fare for passengers from each embark\_town?*

```
df.groupby('embarked').mean()
```

Out[80]:

	survived	pclass	age	sibsp	parch	fare	adult_male	alone
embarked								
C	0.553571	1.886905	30.814769	0.386905	0.363095	59.954144	0.535714	0.5059
Q	0.394737	2.907895	28.089286	0.421053	0.171053	13.348741	0.473684	0.7500
S	0.337481	2.349922	29.444394	0.572317	0.413686	27.109647	0.636081	0.6096

In [81]: *#the average fare for male vs female passengers from each embark\_town*

```
df.groupby('sex').mean()
```

Out[81]:

	survived	pclass	age	sibsp	parch	fare	adult_male	alone
sex								
female	0.742038	2.159236	27.915709	0.694268	0.649682	44.479818	0.000000	0.401274
male	0.189565	2.387826	30.728252	0.429565	0.236522	25.585462	0.930435	0.713043

In [85]: *#Show the summary statistics of data*

```
summary_stats = passengers_df.describe()
print(summary_stats)
```

	ClientID	net worth	years
count	5.000000	5.000000	5.000000
mean	3.000000	31.400000	4.400000
std	1.581139	6.841053	1.516575
min	1.000000	22.000000	2.000000
25%	2.000000	28.000000	4.000000
50%	3.000000	32.000000	5.000000
75%	4.000000	35.000000	5.000000
max	5.000000	40.000000	6.000000

In [ ]: