

ML Project on Employee Dataset

```
import pandas as pd
import numpy as np

df=pd.read_csv('Employee_Salary_Dataset.csv')
print(df)
```

	ID	Experience_Years	Age	Gender	Salary
0	1	5	28	Female	250000
1	2	1	21	Male	50000
2	3	3	23	Female	170000
3	4	2	22	Male	25000
4	5	1	17	Male	10000
5	6	25	62	Male	5001000
6	7	19	54	Female	800000
7	8	2	21	Female	9000
8	9	10	36	Female	61500
9	10	15	54	Female	650000
10	11	4	26	Female	250000
11	12	6	29	Male	1400000
12	13	14	39	Male	6000050
13	14	11	40	Male	220100
14	15	2	23	Male	7500
15	16	4	27	Female	87000
16	17	10	34	Female	930000
17	18	15	54	Female	7900000
18	19	2	21	Male	15000
19	20	10	36	Male	330000
20	21	15	54	Male	6570000
21	22	4	26	Male	25000
22	23	5	29	Male	6845000
23	24	1	21	Female	6000
24	25	4	23	Female	8900
25	26	3	22	Female	20000
26	27	1	18	Male	3000
27	28	27	62	Female	10000000
28	29	19	54	Female	5000000
29	30	2	21	Female	6100
30	31	10	34	Male	80000
31	32	15	54	Male	900000
32	33	20	55	Female	1540000
33	34	19	53	Female	9300000
34	35	16	49	Male	7600000

```
df.head()
```

	ID	Experience_Years	Age	Gender	Salary
0	1	5	28	Female	250000
1	2	1	21	Male	50000

2	3	3	23	Female	170000
3	4	2	22	Male	25000
4	5	1	17	Male	10000

```
print(df.columns)
print(df.shape)
```

```
Index(['ID', 'Experience_Years', 'Age', 'Gender', 'Salary'],
      dtype='object')
(35, 5)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 35 entries, 0 to 34
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    35 non-null    int64
1   Experience_Years      35 non-null    int64
2   Age                  35 non-null    int64
3   Gender               35 non-null    object
4   Salary               35 non-null    int64
dtypes: int64(4), object(1)
memory usage: 1.5+ KB
```

```
#percentage of null valules
(df.isnull().sum()/len(df))*100
```

ID	0.0
Experience_Years	0.0
Age	0.0
Gender	0.0
Salary	0.0

```
dtype: float64
```

```
#removes duplicates
df=df[~df.duplicated()]
df.shape
```

```
(35, 5)
```

```
df.isnull().sum()
df.dtypes
```

ID	int64
Experience_Years	int64
Age	int64
Gender	object
Salary	int64

```
dtype: object
```

```

#mode
df['Age'].mode()

0    54
Name: Age, dtype: int64

#mean
df['Salary'].fillna(df['Salary'].mean())
df['SalaryExperience']=df['Experience_Years'].fillna(df['Experience_Ye
ars'].mean())
print(df.isnull().sum())

ID          0
Experience_Years  0
Age          0
Gender       0
Salary       0
SalaryExperience  0
dtype: int64

df.shape

(35, 6)

df.head()
# and condition
df1=df[(df['Salary']>20000) & (df['Salary']>80000)]
df1.head()

   ID  Experience_Years  Age  Gender  Salary  SalaryExperience
0   1                 5   28  Female   250000                5
2   3                 3   23  Female   170000                3
5   6                25   62   Male  5001000               25
6   7                19   54  Female   800000               19
9  10                15   54  Female   650000               15

df.shape

(35, 6)

df2=df[np.logical_and(df['Salary']>20000,df['Salary']<90000)]
df2.shape

(6, 6)

df.shape

(35, 6)

df.query('Salary>20000 and Salary <80000')
df

```

	ID	Experience_Years	Age	Gender	Salary	SalaryExperience
0	1	5	28	Female	250000	5
1	2	1	21	Male	50000	1
2	3	3	23	Female	170000	3
3	4	2	22	Male	25000	2
4	5	1	17	Male	10000	1
5	6	25	62	Male	5001000	25
6	7	19	54	Female	800000	19
7	8	2	21	Female	9000	2
8	9	10	36	Female	61500	10
9	10	15	54	Female	650000	15
10	11	4	26	Female	250000	4
11	12	6	29	Male	1400000	6
12	13	14	39	Male	6000050	14
13	14	11	40	Male	220100	11
14	15	2	23	Male	7500	2
15	16	4	27	Female	87000	4
16	17	10	34	Female	930000	10
17	18	15	54	Female	7900000	15
18	19	2	21	Male	15000	2
19	20	10	36	Male	330000	10
20	21	15	54	Male	6570000	15
21	22	4	26	Male	25000	4
22	23	5	29	Male	6845000	5
23	24	1	21	Female	6000	1
24	25	4	23	Female	8900	4
25	26	3	22	Female	20000	3
26	27	1	18	Male	3000	1
27	28	27	62	Female	10000000	27
28	29	19	54	Female	5000000	19
29	30	2	21	Female	6100	2
30	31	10	34	Male	80000	10
31	32	15	54	Male	900000	15
32	33	20	55	Female	1540000	20
33	34	19	53	Female	9300000	19
34	35	16	49	Male	7600000	16

```

df=pd.read_csv('Employee_Salary_Dataset.csv')
def function_name(x):
    if x=='Male':
        x='m'
    elif x=='Female':
        x='f'
    else:
        x='other'
    return x
df["Gender1"]=df['Gender'].apply(function_name)
pd.set_option('display.max_rows',None)
pd.set_option('display.max_columns',None)
print(df.head())

```

	ID	Experience_Years	Age	Gender	Salary	Gender1	
0	1		5	28	Female	250000	f
1	2		1	21	Male	50000	m
2	3		3	23	Female	170000	f
3	4		2	22	Male	25000	m
4	5		1	17	Male	10000	m

#bouns

```
df['bouns']=df['Salary']*10/100
df.head()
```

	ID	Experience_Years	Age	Gender	Salary	Gender1	bouns	
0	1		5	28	Female	250000	f	25000.0
1	2		1	21	Male	50000	m	5000.0
2	3		3	23	Female	170000	f	17000.0
3	4		2	22	Male	25000	m	2500.0
4	5		1	17	Male	10000	m	1000.0

```
df['next_year_salary']=df['bouns']+df['Salary']
df.head()
```

	ID	Experience_Years	Age	Gender	Salary	Gender1	bouns	\
0	1		5	28	Female	250000	f	25000.0
1	2		1	21	Male	50000	m	5000.0
2	3		3	23	Female	170000	f	17000.0
3	4		2	22	Male	25000	m	2500.0
4	5		1	17	Male	10000	m	1000.0

	next_year_salary
0	275000.0
1	55000.0
2	187000.0
3	27500.0
4	11000.0

```
df.dtypes
```

```
ID          int64
Experience_Years  int64
Age           int64
Gender        object
Salary        int64
Gender1       object
bouns        float64
next_year_salary float64
dtype: object
```

```
df.head()
```

	ID	Experience_Years	Age	Gender	Salary	Gender1	bouns	\
0	1		5	28	Female	250000	f	25000.0

1	2	1	21	Male	50000	m	5000.0
2	3	3	23	Female	170000	f	17000.0
3	4	2	22	Male	25000	m	2500.0
4	5	1	17	Male	10000	m	1000.0

```

next_year_salary
0      275000.0
1      55000.0
2     187000.0
3      27500.0
4      11000.0

```

```

#revers columns names
print(df.columns[::-1])

```

```

Index(['next_year_salary', 'bouns', 'Gender1', 'Salary', 'Gender',
      'Age',
      'Experience_Years', 'ID'],
      dtype='object')

```

```
df.columns
```

```

Index(['ID', 'Experience_Years', 'Age', 'Gender', 'Salary', 'Gender1',
      'bouns',
      'next_year_salary'],
      dtype='object')

```

```
#row reverse
```

```
df[::-1]
df.head()
```

	ID	Experience_Years	Age	Gender	Salary	Gender1	bouns	\
0	1	5	28	Female	250000	f	25000.0	
1	2	1	21	Male	50000	m	5000.0	
2	3	3	23	Female	170000	f	17000.0	
3	4	2	22	Male	25000	m	2500.0	
4	5	1	17	Male	10000	m	1000.0	

```

next_year_salary
0      275000.0
1      55000.0
2     187000.0
3      27500.0
4      11000.0

```

```
df=pd.pivot_table(df,index='ID',columns='Gender',aggfunc='count')
df.head()
```

	Age	Experience_Years	Gender1	Salary
bouns \				
Gender	Female	Male	Female	Male
	Female	Male	Female	Male

		Female	Male						
ID									
1		1.0	NaN		1.0	NaN		1.0	NaN
1.0	NaN								
2		NaN	1.0		NaN	1.0		NaN	1.0
NaN	1.0								
3		1.0	NaN		1.0	NaN		1.0	NaN
1.0	NaN								
4		NaN	1.0		NaN	1.0		NaN	1.0
NaN	1.0								
5		NaN	1.0		NaN	1.0		NaN	1.0
NaN	1.0								

next_year_salary		
Gender	Female	Male
ID		
1	1.0	NaN
2	NaN	1.0
3	1.0	NaN
4	NaN	1.0
5	NaN	1.0

```
df=pd.read_csv('Employee_Salary_Dataset.csv')
df.columns
```

```
Index(['ID', 'Experience_Years', 'Age', 'Gender', 'Salary'],
      dtype='object')
```

```
df1=df.groupby('Experience_Years')['Salary'].sum()
print(df1)
```

Experience_Years	
1	69000
2	62600
3	190000
4	370900
5	7095000
6	1400000
10	1401500
11	220100
14	6000050
15	16020000
16	7600000
19	15100000
20	1540000
25	5001000
27	10000000

Name: Salary, dtype: int64

```
df1=df.groupby('Experience_Years')
['Salary'].agg(['mean','sum','count'])
df1
```

	mean	sum	count
Experience_Years			
1	1.725000e+04	69000	4
2	1.252000e+04	62600	5
3	9.500000e+04	190000	2
4	9.272500e+04	370900	4
5	3.547500e+06	7095000	2
6	1.400000e+06	1400000	1
10	3.503750e+05	1401500	4
11	2.201000e+05	220100	1
14	6.000050e+06	6000050	1
15	4.005000e+06	16020000	4
16	7.600000e+06	7600000	1
19	5.033333e+06	15100000	3
20	1.540000e+06	1540000	1
25	5.001000e+06	5001000	1
27	1.000000e+07	10000000	1