

```
In [27]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as se
```

```
In [28]: sp=pd.read_csv("/home/student/Desktop/Employee_Salary_Dataset.csv")
```

```
In [4]: sp.min()
```

```
Out[4]: ID          1
Experience_Years    1
Age                17
Gender              Female
Salary             3000
dtype: object
```

```
In [5]: sp.max()
```

```
Out[5]: ID          35
Experience_Years    27
Age                62
Gender              Male
Salary             10000000
dtype: object
```

```
In [6]: sp.std()
```

```
/tmp/ipykernel_2740/2171739191.py:1: FutureWarning: The default value of numeric_only in DataFrame.std is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated. Select only valid columns or specify the value of numeric_only to silence this warning.
  sp.std()
```

```
Out[6]: ID          1.024695e+01
Experience_Years    7.552950e+00
Age                1.464355e+01
Salary             3.170124e+06
dtype: float64
```

```
In [7]: sp.mean()
```

```
/tmp/ipykernel_2740/3291234476.py:1: FutureWarning: The default value of numeric_only in DataFrame.mean is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated. Select only valid columns or specify the value of numeric_only to silence this warning.
  sp.mean()
```

```
Out[7]: ID          1.800000e+01
Experience_Years    9.200000e+00
Age                3.548571e+01
Salary             2.059147e+06
dtype: float64
```

In [8]: `sp.mode()`

Out[8]:

	ID	Experience_Years	Age	Gender	Salary
0	1	2.0	54.0	Female	25000.0
1	2	NaN	NaN	NaN	250000.0
2	3	NaN	NaN	NaN	NaN
3	4	NaN	NaN	NaN	NaN
4	5	NaN	NaN	NaN	NaN
5	6	NaN	NaN	NaN	NaN
6	7	NaN	NaN	NaN	NaN
7	8	NaN	NaN	NaN	NaN
8	9	NaN	NaN	NaN	NaN
9	10	NaN	NaN	NaN	NaN
10	11	NaN	NaN	NaN	NaN
11	12	NaN	NaN	NaN	NaN
12	13	NaN	NaN	NaN	NaN
13	14	NaN	NaN	NaN	NaN
14	15	NaN	NaN	NaN	NaN
15	16	NaN	NaN	NaN	NaN
16	17	NaN	NaN	NaN	NaN
17	18	NaN	NaN	NaN	NaN
18	19	NaN	NaN	NaN	NaN
19	20	NaN	NaN	NaN	NaN
20	21	NaN	NaN	NaN	NaN
21	22	NaN	NaN	NaN	NaN
22	23	NaN	NaN	NaN	NaN
23	24	NaN	NaN	NaN	NaN
24	25	NaN	NaN	NaN	NaN
25	26	NaN	NaN	NaN	NaN
26	27	NaN	NaN	NaN	NaN
27	28	NaN	NaN	NaN	NaN
28	29	NaN	NaN	NaN	NaN
29	30	NaN	NaN	NaN	NaN
30	31	NaN	NaN	NaN	NaN
31	32	NaN	NaN	NaN	NaN
32	33	NaN	NaN	NaN	NaN

ID	Experience_Years	Age	Gender	Salary
0	1	19	M	62508.50
1	2	21	F	12506.00
2	3	22	M	42507.25
3	4	25	M	6257.00

```
In [9]: sp.mean(axis=1)[0:4]
```

```
/tmp/ipykernel_2740/2676889982.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.
  sp.mean(axis=1)[0:4]
```

```
Out[9]: 0    62508.50
        1    12506.00
        2    42507.25
        3     6257.00
        dtype: float64
```

```
In [10]: sp.groupby(['Experience_Years'])['Age'].mean()
```

```
Out[10]: Experience_Years
1      19.250000
2      21.600000
3      22.500000
4      25.500000
5      28.500000
6      29.000000
10     35.000000
11     40.000000
14     39.000000
15     54.000000
16     49.000000
19     53.666667
20     55.000000
25     62.000000
27     62.000000
Name: Age, dtype: float64
```

```
In [15]: sp_u = sp.rename(columns={'Experience_Years': 'E_year'}, inplace=False)
```

```
In [16]: (sp_u.groupby(['Age']).E_year.mean())
```

```
Out[16]:
```

```
Age
17      1.000000
18      1.000000
```

In [17]: `sp.head()`

Out[17]:

	ID	E_year	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

In [19]: `sp.std(axis=1)[0:4]`

```
/tmp/ipykernel_2740/2545776365.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.
  sp.std(axis=1)[0:4]
```

Out[19]:

```
0    124994.333900
1     24996.001694
2     84995.167190
3     12495.336570
dtype: float64
```

In [20]: `from sklearn.preprocessing import LabelEncoder`
`le = LabelEncoder()`

In [24]: `sp['Gender'] = le.fit_transform(sp['Gender'])`
`newdf = sp`
`sp`

Out[24]:

	ID	E_year	Age	Gender	Salary
0	1	5	28	0	250000
1	2	1	21	1	50000
2	3	3	23	0	170000
3	4	2	22	1	25000
4	5	1	17	1	10000
5	6	25	62	1	5001000
6	7	19	54	0	800000
7	8	2	21	0	9000
8	9	10	36	0	61500
9	10	15	54	0	650000
10	11	4	26	0	250000

	ID	E_year	Age	Gender	Salary
11	12	6	29	1	1400000
12	13	14	39	1	6000050
13	14	11	40	1	220100
14	15	2	23	1	7500
15	16	4	27	0	87000
16	17	10	34	0	930000
17	18	15	54	0	7900000
18	19	2	21	1	15000
19	20	10	36	1	330000
20	21	15	54	1	6570000
21	22	4	26	1	25000
22	23	5	29	1	6845000
23	24	1	21	0	6000
24	25	4	23	0	8900
25	26	3	22	0	20000
26	27	1	18	1	3000
27	28	27	62	0	10000000
28	29	19	54	0	5000000
29	30	2	21	0	6100
30	31	10	34	1	80000
31	32	15	54	1	900000
32	33	20	55	0	1540000

In [25]: `sp.isnull()`

Out[25]:

	ID	E_year	Age	Gender	Salary
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
5	False	False	False	False	False
6	False	False	False	False	False
7	False	False	False	False	False
8	False	False	False	False	False
9	False	False	False	False	False

	ID	E_year	Age	Gender	Salary
10	False	False	False	False	False
11	False	False	False	False	False
12	False	False	False	False	False
13	False	False	False	False	False
14	False	False	False	False	False
15	False	False	False	False	False
16	False	False	False	False	False
17	False	False	False	False	False
18	False	False	False	False	False
19	False	False	False	False	False
20	False	False	False	False	False
21	False	False	False	False	False
22	False	False	False	False	False
23	False	False	False	False	False
24	False	False	False	False	False
25	False	False	False	False	False
26	False	False	False	False	False
27	False	False	False	False	False
28	False	False	False	False	False
29	False	False	False	False	False
30	False	False	False	False	False
31	False	False	False	False	False
32	False	False	False	False	False
33	False	False	False	False	False

In [29]: `sp.dropna(how='all')`

Out[29]:

	ID	Experience_Years	Age	Gender	Salary
0	1	NaN	28	Female	250000
1	2	1.0	21	Male	50000
2	3	3.0	23	Female	170000
3	4	2.0	22	Male	25000
4	5	1.0	17	Male	10000
5	6	25.0	62	Male	5001000
6	7	19.0	54	Female	800000
7	8	2.0	21	Female	9000

	ID	Experience_Years	Age	Gender	Salary
8	9	10.0	36	Female	61500
9	10	15.0	54	Female	650000
10	11	4.0	26	Female	250000
11	12	6.0	29	Male	1400000
12	13	14.0	39	Male	6000050
13	14	11.0	40	Male	220100
14	15	2.0	23	Male	7500
15	16	4.0	27	Female	87000
16	17	10.0	34	Female	930000
17	18	15.0	54	Female	7900000
18	19	2.0	21	Male	15000
19	20	10.0	36	Male	330000
20	21	15.0	54	Male	6570000
21	22	4.0	26	Male	25000
22	23	5.0	29	Male	6845000
23	24	1.0	21	Female	6000
24	25	4.0	23	Female	8900
25	26	3.0	22	Female	20000
26	27	1.0	18	Male	3000
27	28	27.0	62	Female	10000000
28	29	19.0	54	Female	5000000
29	30	2.0	21	Female	6100
30	31	10.0	34	Male	80000
31	32	15.0	54	Male	900000
32	33	20.0	55	Female	1540000
33	34	10.0	53	Female	9300000

```
In [30]: sp=pd.read_csv("/home/student/Desktop/Iris.csv")
```

```
In [34]: from sklearn import preprocessing
enc = preprocessing.OneHotEncoder()
enc_df= pd.DataFrame(enc.fit_transform(sp[['SepalWidthCm']])).toarray()
enc_df
```

```
Out[34]:
```

	0	1	2	3	4	5	6	7	8	9	...	13	14	15	16	17	18	19	20
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

	0	1	2	3	4	5	6	7	8	9	...	13	14	15	16	17	18	19	20
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0
...
145	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
146	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
147	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
148	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
149	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	...	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

```
In [36]: df_encode = sp_u.join(enc_df)
df_encode
```

Out[36]:

	ID	E_year	Age	Gender	Salary	0	1	2	3	4	...	13	14	15	16	17
0	1	5	28	Female	250000	0.0	0.0	0.0	0.0	0.0	...	0.0	1.0	0.0	0.0	0.0
1	2	1	21	Male	50000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
2	3	3	23	Female	170000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
3	4	2	22	Male	25000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
4	5	1	17	Male	10000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	1.0	0.0	0.0
5	6	25	62	Male	5001000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
6	7	19	54	Female	800000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
7	8	2	21	Female	9000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
8	9	10	36	Female	61500	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
9	10	15	54	Female	650000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
10	11	4	26	Female	250000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	1.0	0.0
11	12	6	29	Male	1400000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
12	13	14	39	Male	6000050	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
13	14	11	40	Male	220100	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
14	15	2	23	Male	7500	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
15	16	4	27	Female	87000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
16	17	10	34	Female	930000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
17	18	15	54	Female	7900000	0.0	0.0	0.0	0.0	0.0	...	0.0	1.0	0.0	0.0	0.0
18	19	2	21	Male	15000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	1.0
19	20	10	36	Male	330000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	1.0
20	21	15	54	Male	6570000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
21	22	4	26	Male	25000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	1.0	0.0
22	23	5	29	Male	6845000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	1.0	0.0	0.0
23	24	1	21	Female	6000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0

	ID	E_year	Age	Gender	Salary	0	1	2	3	4	...	13	14	15	16	17
24	25	4	23	Female	8900	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
25	26	3	22	Female	20000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
26	27	1	18	Male	3000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
27	28	27	62	Female	10000000	0.0	0.0	0.0	0.0	0.0	...	0.0	1.0	0.0	0.0	0.0
28	29	19	54	Female	5000000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
29	30	2	21	Female	6100	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
30	31	10	34	Male	80000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
31	32	15	54	Male	900000	0.0	0.0	0.0	0.0	0.0	...	1.0	0.0	0.0	0.0	0.0
32	33	20	55	Female	1540000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
33	34	19	53	Female	9300000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
34	35	16	49	Male	7600000	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0

```
In [41]: irisSet = (sp['Species']== 'Iris-setosa')
print('Iris-setosa')
print(sp[irisSet].describe())
```

```
Iris-setosa
      Id  SepalLengthCm  SepalWidthCm  PetalLengthCm  Peta
lWidthCm
count  50.00000      50.00000      50.000000      50.000000
50.00000
mean   25.50000        5.00600      3.418000      1.464000
0.24400
std    14.57738        0.35249      0.381024      0.173511
0.10721
min     1.00000        4.30000      2.300000      1.000000
0.10000
25%    13.25000        4.80000      3.125000      1.400000
0.20000
50%    25.50000        5.00000      3.400000      1.500000
0.20000
75%    37.75000        5.20000      3.675000      1.575000
0.30000
max    50.00000        5.80000      4.400000      1.900000
0.60000
```

```
In [42]: irisVer = (sp['Species']== 'Iris-setosa')
print('Iris-setosa')
print(sp[irisVer].describe())
```

Iris-setosa

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.00000	50.00000	50.00000	50.00000
mean	25.50000	5.00600	3.41800	1.46400	0.24400
std	14.57738	0.35249	0.381024	0.173511	0.10721

```
In [44]: irisVir = (sp['Species']=='Iris-setosa')
print('Iris-setosa')
print(sp[irisVir].describe())
```

Iris-setosa

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.00000	50.00000	50.00000	50.00000
mean	25.50000	5.00600	3.41800	1.46400	0.24400
std	14.57738	0.35249	0.381024	0.173511	0.10721
min	1.00000	4.30000	2.30000	1.00000	0.10000
25%	13.25000	4.80000	3.12500	1.40000	0.20000
50%	25.50000	5.00000	3.40000	1.50000	0.20000
75%	37.75000	5.20000	3.67500	1.57500	0.30000
max	50.00000	5.80000	4.40000	1.90000	0.60000

```
In [ ]:
```