SALARY DATA STATISTICS

```
import numpy as np
 In [4]:
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
 In [6]:
         import warnings
         warnings.filterwarnings('ignore')
        salary = pd.read_csv(r"E:\Data Science & AI\Dataset files\Salary_Data.csv")
In [22]:
In [24]: salary.head()
Out[24]:
            YearsExperience Salary
          0
                        1.1
                            39343
                        1.3 46205
          1
          2
                        1.5 37731
          3
                        2.0 43525
          4
                        2.2 39891
In [26]:
         salary.shape
Out[26]: (30, 2)
In [28]:
        salary.columns
Out[28]: Index(['YearsExperience', 'Salary'], dtype='object')
In [30]:
         salary.isnull().sum()
Out[30]: YearsExperience
                             0
          Salary
          dtype: int64
In [32]: salary.dtypes
                             float64
Out[32]: YearsExperience
                               int64
          Salary
          dtype: object
In [38]: x = salary.drop('YearsExperience',axis=1)
         y = salary['Salary']
In [48]: from sklearn.model_selection import train_test_split
         x_train, x_test, y_train, y_test=train_test_split(x,y,random_state=1234,test_siz
In [50]: x_train.shape,x_test.shape
```

```
In [52]: y_train.shape,y_test.shape
Out[52]: ((21,), (9,))
In [56]: salary.shape
Out[56]: (30, 2)
In [58]: x_train
Out[58]: Salary
         13 57081
         22 101302
         24 109431
         0 39343
         2 37731
         27 112635
         26 116969
         18 81363
         5 56642
        16 66029
         25 105582
        11 55794
         9 57189
        17 83088
        29 121872
        20 91738
         12 56957
        21 98273
         6 60150
        19 93940
         15 67938
```

Out[50]: ((21, 1), (9, 1))

In [60]: y_train

```
Out[60]: 13
               57081
         22 101302
         24
             109431
         0
               39343
         2
               37731
         27
               112635
         26
               116969
         18
               81363
         5
               56642
         16
               66029
         25
               105582
         11
              55794
         9
               57189
         17
               83088
         29
             121872
         20
               91738
         12
               56957
         21
                98273
                60150
         6
         19
                93940
                67938
         15
         Name: Salary, dtype: int64
In [62]: x_test
Out[62]:
             Salary
          7
              54445
              63218
         10
              39891
          4
              46205
         28 122391
              64445
          3
              43525
         23
            113812
         14
             61111
In [64]: y_test
Out[64]: 7
                54445
         10
                63218
         4
               39891
         1
               46205
         28
               122391
         8
               64445
         3
               43525
         23
               113812
                61111
         Name: Salary, dtype: int64
In [66]: x_train.ndim
```

```
In [68]: from sklearn.linear_model import LinearRegression
         LR = LinearRegression()
         LR.fit(x_train,y_train)
Out[68]:
             LinearRegression 🕘
         LinearRegression()
In [70]: y_predications = LR.predict(x_test)
In [72]: y_predications
Out[72]: array([ 54445., 63218., 39891., 46205., 122391., 64445., 43525.,
                113812., 61111.])
In [74]: y_test.shape,y_predications.shape
Out[74]: ((9,), (9,))
In [76]: x_test
Out[76]:
              Salary
              54445
          7
         10
              63218
              39891
          4
              46205
          1
         28 122391
          8
             64445
          3
             43525
         23 113812
             61111
         14
In [80]: x_test.iloc[0]
         x_test.iloc[0].values
Out[80]: array([54445], dtype=int64)
In [86]: LR.predict([x_test.iloc[0].values,x_test.iloc[1].values])
Out[86]: array([54445., 63218.])
In [88]: ip1 = [5]
         LR.predict([ip1])
```

Out[66]: 2

Out[88]: array([5.])

```
In [92]: x_test.shape,y_test.shape,y_predications.shape
Out[92]: ((9, 1), (9,), (9,))
In [96]: test_data=x_test
          test_data['y_actual']=y_test
          test_data['y_predictions']=y_predications
          test_data
Out[96]:
               Salary y_actual y_predictions
            7
                54445
                         54445
                                     54445.0
           10
                63218
                         63218
                                     63218.0
                39891
                         39891
                                     39891.0
                46205
                         46205
                                     46205.0
           28 122391
                        122391
                                    122391.0
                64445
                         64445
                                     64445.0
               43525
                        43525
                                     43525.0
            3
           23 113812
                        113812
                                    113812.0
                61111
                         61111
                                     61111.0
           14
In [102...
          print(y_test.values[:5])
         [ 54445 63218 39891 46205 122391]
In [106...
         print(y_predications[:5])
         [ 54445. 63218. 39891. 46205. 122391.]
In [108...
          from sklearn.metrics import r2_score,mean_squared_error
In [114...
          R2=r2_score(y_test,y_predications)
          MSE=mean_squared_error(y_test,y_predications)
          \#MSE^{**}(1/2)
          RMSE=np.sqrt(MSE)
          #accuracy_score(y_test,y_predictions) # it is a regression tech
          print("R-sqaure:",R2)
          print("MSE:",MSE)
          print("RMSE:",RMSE)
         R-sqaure: 1.0
         MSE: 4.117521271375071e-23
         RMSE: 6.41679146565873e-12
In [120...
          s=0
          for i in range(len(y test)):
              v1=y_test.values[i]-y_predications[i]
              v2=v1**2
              s=s+v2
          print(s/len(y_test))
         4.117521271375071e-23
```

```
In [122...
          LR.coef_
          print("The coefficent of Years_of_experience is:",LR.coef_)
         The coefficent of Years_of_experience is: [1.]
In [124...
         LR.intercept_
Out[124... 1.4551915228366852e-11
In [126... x_train.columns
Out[126... Index(['Salary'], dtype='object')
In [138...
         from sklearn.feature_selection import VarianceThreshold
          vt=VarianceThreshold(threshold=0)
          # Threshold variance value
          # we want to drop the feaure based on threshold
          vt.fit(salary)
Out[138...
                VarianceThreshold
          VarianceThreshold(threshold=0)
In [140...
           dir(vt)
```

```
Out[140... ['__abstractmethods__',
             '__annotations__',
'__class__',
              '__delattr__',
               __dict__',
             '__dir__',
'__doc__',
             '__format__',
             ___ge__',
               __getattribute__',
             '__getstate__',
             __gt__',
'__hash__',
'__init__',
               __init_subclass___',
             __le__',
'__lt___',
               __module__',
              '__ne__',
             '__new__',
               __reduce__',
             '__reduce_ex__',
             '__repr__',
'__setattr__',
'__setstate__',
              '__sizeof__',
              '__sklearn_clone__',
             '__str__',
'__subclasshook__',
             '__weakref__',
              '_abc_impl',
              _____'_build_request_for_signature',
              '_check_feature_names',
             '_check_n_features',
              '_doc_link_module',
              '_doc_link_template',
             '_doc_link_url_param_generator',
              '_get_default_requests',
              _get_doc_link',
              '_get_metadata_request',
              '_get_param_names',
              '_get_support_mask',
              '_get_tags',
             '_more_tags',
             '_parameter_constraints',
              ___
'_repr_html_',
              '_repr_html_inner',
              '_repr_mimebundle_',
              '_sklearn_auto_wrap_output_keys',
             ____
'_transform',
              '_validate_data',
             '_validate_params',
              'feature_names_in_',
              'fit',
             'fit_transform',
              'get_feature_names_out',
              'get_metadata_routing',
              'get_params',
              'get_support',
```

```
'inverse_transform',
            'n_features_in_',
            'set_output',
            'set_params',
            'threshold',
            'transform',
            'variances_']
In [142...
          vt.variances_
          # 300 is first column variance (T)
          # 1.25 is second column variance (T)
          # 30 is column varaince (T)
          # 0 is fourth column variance (F)
Out[142... array([7.78515556e+00, 8.46600000e+04])
In [144... vt.get_support()
Out[144... array([ True, True])
In [146...
          vt.get_support()
Out[146... array([ True, True])
          vt.get_params()
In [148...
Out[148... {'threshold': 0}
          vt.threshold
In [150...
Out[150... 0
In [154...
          cols=vt.get_feature_names_out()
          # the above syntax gives the column names
          # These fetaure only we want include
          salary[cols]
```

	YearsExperience	Salary
0	1.1	39343
1	1.3	46205
2	1.5	37731
3	2.0	43525
4	2.2	39891
5	2.9	56642
6	3.0	60150
7	3.2	54445
8	3.2	64445
9	3.7	57189
10	3.9	63218
11	4.0	55794
12	4.0	56957
13	4.1	57081
14	4.5	61111
15	4.9	67938
16	5.1	66029
17	5.3	83088
18	5.9	81363
19	6.0	93940
20	6.8	91738
21	7.1	98273
22	7.9	101302
23	8.2	113812
24	8.7	109431
25	9.0	105582
26	9.5	116969
27	9.6	112635
28	10.3	122391
29	10.5	121872

salary=pd.read_csv(r"E:\Data Science & AI\Dataset files\Salary_Data.csv") salary.head()

 $\textbf{from} \ \, \textbf{sklearn.feature_selection} \ \, \textbf{import} \ \, \textbf{VarianceThreshold}$

```
vt=VarianceThreshold(threshold=0)
### Make sure before fitting the dataframe , do not include output column
x=salary.drop('YearsExperience',axis=1)
# x it self a data frame
vt.fit(x)
vt.variances_
vt.get_support()
cols=vt.get_feature_names_out()
x[cols]
```

- 39343
- 46205
- 37731
- 43525
- 39891
- 56642
- 60150
- 54445
- 64445
- 57189
- 63218
- 55794
- 56957
- 57081
- 61111
- 67938
- 66029
- 83088
- 81363
- 93940
- 91738
- 98273
- 101302
- 113812
- 109431
- 105582
- 116969
- 112635
- 122391
- 121872

OLS Regression Results

Dep. Variable:	Salary	R-squared (uncentered):	1.000
Model:	OLS	Adj. R-squared (uncentered):	1.000
Method:	Least Squares	F-statistic:	1.583e+32
Date:	Fri, 20 Sep 2024	Prob (F-statistic):	1.82e-310
Time:	21:48:30	Log-Likelihood:	479.24
No. Observations:	21	AIC:	-956.5
Df Residuals:	20	BIC:	-955.4
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Salary	1.0000	7.95e-17	1.26e+16	0.000	1.000	1.000

Omnibus: 1.354 **Durbin-Watson:** 0.163 **Prob(Omnibus):** 0.508 **Jarque-Bera (JB):** 1.139 **Skew:** 0.522 **Prob(JB):** 0.566 Kurtosis: 2.539 Cond. No. 1.00

Notes:

In []:

- [1] R² is computed without centering (uncentered) since the model does not contain a constant.
- [2] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [162...
          import pickle
          pickle.dump(LR,open('YearsExperience_model.pkl','wb'))
In [164...
          # loading model to comapare the result
          model=pickle.load(open('YearsExperience_model.pkl','rb'))
          model
Out[164...
               LinearRegression
          LinearRegression()
```