```
import pandas as pd
import numpy as np
```

data = pd.read_csv("Engineering_graduate_salary.csv")

data

	ID	Gender	DOB	10percentage	10board	12graduation	12percentage	:
0	604399	f	1990- 10-22	87.80	cbse	2009	84.00	
1	988334	m	1990- 05-15	57.00	cbse	2010	64.50	
2	301647	m	1989- 08-21	77.33	maharashtra state board,pune	2007	85.17	i d
3	582313	m	1991- 05-04	84.30	cbse	2009	86.00	
4	339001	f	1990- 10-30	82.00	cbse	2008	75.00	
2993	103174	f	1989- 04-17	75.00	0	2005	73.00	
2994	352811	f	1991- 07-22	84.00	state board	2008	77.00	
2995	287070	m	1988- 11-24	91.40	bsemp	2006	65.56	
2996	317336	m	1988- 08-25	88.64	karnataka education board	2006	65.16	ka e
2997	993701	m	1992- 05-27	77.00	state board	2009	75.50	

2998 rows × 34 columns

data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2998 entries, 0 to 2997
Data columns (total 34 columns):
```

Column Non-Null Count Dtype
--- --- 2998 non-null int64

1	Gender	2998	non-null	object
2	DOB	2998	non-null	object
3	10percentage	2998	non-null	float64
4	10board	2998	non-null	object
5	12graduation	2998	non-null	int64
6	12percentage	2998	non-null	float64
7	12board	2998	non-null	object
8	CollegeID	2998	non-null	int64
9	CollegeTier	2998	non-null	int64
10	Degree	2998	non-null	object
11	Specialization	2998	non-null	object
12	collegeGPA	2998	non-null	float64
13	CollegeCityID	2998	non-null	int64
14	CollegeCityTier	2998	non-null	int64
15	CollegeState	2998	non-null	object
16	GraduationYear	2998	non-null	int64
17	English	2998	non-null	int64
18	Logical	2998	non-null	int64
19	Quant	2998	non-null	int64
20	Domain	2998	non-null	float64
21	ComputerProgramming	2998	non-null	int64
22	ElectronicsAndSemicon	2998	non-null	int64
23	ComputerScience	2998	non-null	int64
24	MechanicalEngg	2998	non-null	int64
25	ElectricalEngg	2998	non-null	int64
26	TelecomEngg	2998	non-null	int64
27	CivilEngg	2998	non-null	int64
28	conscientiousness	2998	non-null	float64
29	agreeableness	2998	non-null	float64
30	extraversion	2998	non-null	float64
31	nueroticism	2998	non-null	float64
32	openess_to_experience	2998	non-null	float64
33	Salary	2998	non-null	int64
dtyp	es: float64(9), int64(1	8), ob	oject(7)	

dtypes: float64(9), int64(18), object(7)
memory usage: 796.5+ KB

data.head()

	ID	Gender	DOB	10percentage	10board	12graduation	12percentage	12bo
0	604399	f	1990- 10-22	87.80	cbse	2009	84.00	С
1	988334	m	1990- 05-15	57.00	cbse	2010	64.50	С
2	301647	m	1989- 08-21	77.33	maharashtra state board,pune	2007	85.17	amra division bo
3	582313	m	1991- 05-04	84.30	cbse	2009	86.00	С
4	339001	f	1990- 10-30	82.00	cbse	2008	75.00	С

5 rows × 34 columns

d=data.drop(['ID','10board','12board','Degree','CollegeState','DOB','12graduation','Gender
d.head()

	10percentage	12percentage	CollegeID	CollegeTier	collegeGPA	CollegeCityID	Co
0	87.80	84.00	6920	1	73.82	6920	
1	57.00	64.50	6624	2	65.00	6624	
2	77.33	85.17	9084	2	61.94	9084	
3	84.30	86.00	8195	1	80.40	8195	
4	82.00	75.00	4889	2	64.30	4889	

5 rows × 25 columns

d.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2998 entries, 0 to 2997
Data columns (total 25 columns):

	Calumna (Cotal 25 Columna			Dhum
#	Column	Non-I	Null Count	Dtype
0	10percentage		non-null	float64
1	12percentage	2998	non-null	float64
2	CollegeID		non-null	int64
3	CollegeTier	2998	non-null	int64
4	collegeGPA	2998	non-null	float64
5	CollegeCityID	2998	non-null	int64
6	CollegeCityTier	2998	non-null	int64
7	GraduationYear	2998	non-null	int64
8	English	2998	non-null	int64
9	Logical	2998	non-null	int64
10	Quant	2998	non-null	int64
11	Domain	2998	non-null	float64
12	ComputerProgramming	2998	non-null	int64
13	ElectronicsAndSemicon	2998	non-null	int64
14	ComputerScience	2998	non-null	int64
15	MechanicalEngg	2998	non-null	int64
16	ElectricalEngg	2998	non-null	int64
17	TelecomEngg	2998	non-null	int64
18	CivilEngg	2998	non-null	int64
19	conscientiousness	2998	non-null	float64
20	agreeableness	2998	non-null	float64
21	extraversion	2998	non-null	float64
22	nueroticism	2998	non-null	float64
23	openess_to_experience	2998	non-null	float64
24	Salary	2998	non-null	int64
dtype	es: float64(9), int64(1	6)		

dtypes: float64(9), int64(16)
memory usage: 585.7 KB

X,Y=d.iloc[:,:-1],d.iloc[:,[-1]]

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	10percentage	12percentage	CollegeID	CollegeTier	collegeGPA	CollegeCityID
0	87.80	84.00	6920	1	73.82	6920
1	57.00	64.50	6624	2	65.00	6624
2	77.33	85.17	9084	2	61.94	9084
3	84.30	86.00	8195	1	80.40	8195
4	82.00	75.00	4889	2	64.30	4889
2993	75.00	73.00	1263	2	70.00	1263
2994	84.00	77.00	9481	2	75.20	9481
2995	91.40	65.56	547	2	73.19	547
2996	88.64	65.16	1629	2	74.81	1629
2997	77.00	75.50	1111	2	69.30	1111

2998 rows × 24 columns

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```
Salary
```

```
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
X=scaler.fit_transform(X)
```

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Y=Y/10000

4 200000

Double-click (or enter) to edit

```
from sklearn.model_selection import train_test_split

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X_train,X_test,Y_train,Y_test= train_test_split(X,Y,test_size=1/4,random_state=0)

2006 530000

print(X_train.shape,X_test.shape,Y_train.shape,Y_test.shape)

(2248, 24) (750, 24) (2248, 1) (750, 1)
```

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Y_train.head()

	Salary
487	33.0
2820	25.0
1093	43.0
438	24.0
748	36.0

(750, 1)

```
from sklearn.linear_model import LinearRegression
regressor=LinearRegression()
regressor.fit(X_train,Y_train)
    LinearRegression()
y_pred.shape
```

y_pred

```
array([[ 3.38660895e-03],
       [ 1.63434092e-03],
       [ 3.06641566e-03],
       [ 8.82524739e-04],
       [ 5.03246360e-03],
       [ 2.59535445e-03],
       [ 3.62880130e-03].
       [ 2.80408094e-03],
       [ 3.01973208e-03],
       [ 4.40253356e-03],
         3.38848008e-03],
       [ 4.68101396e-03],
       [ 3.71957116e-03],
       [ 1.62604984e-03],
         2.45631900e-03],
       [ 3.21095260e-03],
       [ 3.17886779e-03],
         3.00002991e-031.
       [ 2.22359135e-03],
       [ 2.60966533e-03],
         3.13928124e-03],
       [ 2.83306883e-03],
       [ 4.19182434e-03].
       [ 4.97112608e-03],
       [ 2.62268031e-03],
       [ 3.84087067e-03],
       [ 2.32565801e-03],
       [ 1.68550577e-03],
       [ 2.10079178e-03],
       [ 3.30297959e-03],
       [ 4.49853591e-03],
       [ 1.97993675e-03],
       [ 3.95819843e-03],
       [ 3.04621652e-03],
       [ 2.57195494e-03],
       [ 4.97858135e-03],
       [ 4.06063303e-03],
       [ 3.03281846e-03],
       [ 2.23685134e-03],
       [ 2.56726036e-03],
       [ 3.06749144e-03],
       [ 4.26900997e-03],
         7.51625160e-01],
       [ 2.93336693e-03],
       [ 1.91409860e-03],
         2.14098401e-03],
       [ 3.20624326e-03],
       [ 2.74087382e-03],
         3.83693201e-03],
       [ 4.28316841e-03],
       [ 2.12103047e-03],
       [ 2.76773606e-03],
       [ 2.85177305e-03],
       [ 4.68733181e-03],
         3.78334553e-03],
       [ 3.92424284e-03],
```

```
[ 3.60090684e-03],
             [ / KOO//OO1__OZ]
y pred=regressor.predict(X test)
y_pred.shape
     (750, 1)
y_pred
     array([[ 3.38660895e+01],
             [ 1.63434092e+01],
             [ 3.06641566e+01],
             [ 8.82524739e+00],
             [ 5.03246360e+01],
             [ 2.59535445e+01],
              3.62880130e+011.
             [ 2.80408094e+01],
             [ 3.01973208e+01],
             [ 4.40253356e+01],
             [ 3.38848008e+01],
             [ 4.68101396e+01],
              3.71957116e+01],
             [ 1.62604984e+01],
             [ 2.45631900e+01],
             [ 3.21095260e+01],
              3.17886779e+01],
             [ 3.00002991e+01],
              2.22359135e+01],
             [ 2.60966533e+01],
             [ 3.13928124e+01],
             [ 2.83306883e+01],
              4.19182434e+011.
             [ 4.97112608e+01],
             [ 2.62268031e+01],
              3.84087067e+01],
             [ 2.32565801e+01],
             [ 1.68550577e+01],
              2.10079178e+01],
             [ 3.30297959e+01],
             [ 4.49853591e+01],
              1.97993675e+01],
             [ 3.95819843e+01],
             [ 3.04621652e+01],
              2.57195494e+01],
             [ 4.97858135e+01],
             [ 4.06063303e+01],
              3.03281846e+01],
             [ 2.23685134e+01],
             [ 2.56726036e+01],
             [ 3.06749144e+01],
             [ 4.26900997e+01],
             [ 7.51625160e+03],
              2.93336693e+01],
             [ 1.91409860e+01],
             [ 2.14098401e+01],
             [ 3.20624326e+01],
```

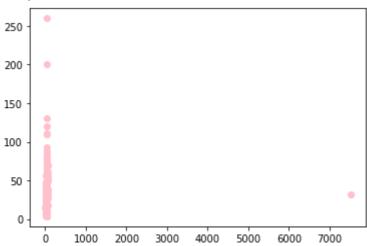


```
[ 2.74087382e+01],
[ 3.83693201e+01],
[ 4.28316841e+01],
[ 2.12103047e+01],
[ 2.76773606e+01],
[ 2.85177305e+01],
[ 4.68733181e+01],
[ 3.78334553e+01],
[ 3.92424284e+01],
[ 3.60090684e+01],
[ 4.60044091e+011.
```

import matplotlib.pyplot as plt

plt.scatter(y_pred,Y_test,color='pink')

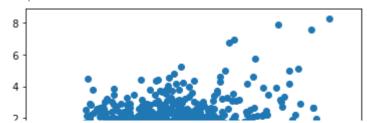
<matplotlib.collections.PathCollection at 0x7f7544017d50>



Double-click (or enter) to edit

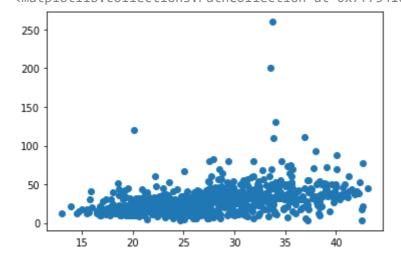
```
from sklearn.decomposition import PCA
pca=PCA(n_components=2)
X_pca=pca.fit_transform(X_train)
plt.scatter(X_pca[:,0],X_pca[:,1])
```

<matplotlib.collections.PathCollection at 0x7f7541e36e10>



from sklearn.svm import SVR
model=SVR()
model.fit(X_train,Y_train)
y_pred=model.predict(X_test)
plt.scatter(y_pred,Y_test)

/usr/local/lib/python3.7/dist-packages/sklearn/utils/validation.py:993: DataConversic
 y = column_or_1d(y, warn=True)
<matplotlib.collections.PathCollection at 0x7f7541e430d0>



score=model.score(X_train,Y_train)
score

0.1084330628881951

from sklearn.ensemble import RandomForestRegressor
rf=RandomForestRegressor()
rf.fit(X_train,Y_train)
y_pred=rf.predict(X_test)
score=rf.score(X_train,Y_train)
score

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: DataConversionWarning
This is separate from the ipykernel package so we can avoid doing imports until
0.8622259002906525

→

plt.scatter(y_pred,Y_test)

<matplotlib.collections.PathCollection at 0x7f754123db50>

