#### In [1]:

#### !pip install panda

Requirement already satisfied: panda in ./anaconda3/lib/python3.11/s ite-packages (0.3.1)

Requirement already satisfied: setuptools in ./anaconda3/lib/python 3.11/site-packages (from panda) (68.0.0)

Requirement already satisfied: requests in ./anaconda3/lib/python3.1 1/site-packages (from panda) (2.31.0)

Requirement already satisfied: charset-normalizer<4,>=2 in ./anacond a3/lib/python3.11/site-packages (from requests->panda) (2.0.4)

Requirement already satisfied: idna<4,>=2.5 in ./anaconda3/lib/pytho n3.11/site-packages (from requests->panda) (3.4)

Requirement already satisfied: urllib3<3,>=1.21.1 in ./anaconda3/lib/python3.11/site-packages (from requests->panda) (1.26.16)

Requirement already satisfied: certifi>=2017.4.17 in ./anaconda3/lib/python3.11/site-packages (from requests->panda) (2023.7.22)

#### In [2]:

#### !pip install numpy

Requirement already satisfied: numpy in ./anaconda3/lib/python3.11/s ite-packages (1.24.3)

#### In [3]:

### import matplotlib.pyplot as plt

#### In [4]:

```
import pandas as pd
import numpy as np
from sklearn.linear model import LinearRegression
```

#### In [5]:

```
data=pd.read csv("Salary Data.csv")
```

# In [6]:

data

# Out[6]:

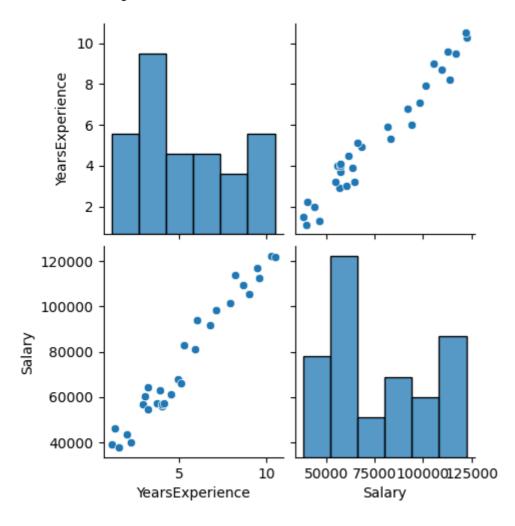
	YearsExperience	Salary
0	1.1	39343.0
1	1.3	46205.0
2	1.5	37731.0
3	2.0	43525.0
4	2.2	39891.0
5	2.9	56642.0
6	3.0	60150.0
7	3.2	54445.0
8	3.2	64445.0
9	3.7	57189.0
10	3.9	63218.0
11	4.0	55794.0
12	4.0	56957.0
13	4.1	57081.0
14	4.5	61111.0
15	4.9	67938.0
16	5.1	66029.0
17	5.3	83088.0
18	5.9	81363.0
19	6.0	93940.0
20	6.8	91738.0
21	7.1	98273.0
22	7.9	101302.0
23	8.2	113812.0
24	8.7	109431.0
25	9.0	105582.0
26	9.5	116969.0
27	9.6	112635.0
28	10.3	122391.0
29	10.5	121872.0

# In [60]:

import seaborn as sns
sns.pairplot(data)

# Out[60]:

<seaborn.axisgrid.PairGrid at 0x7f500e381750>

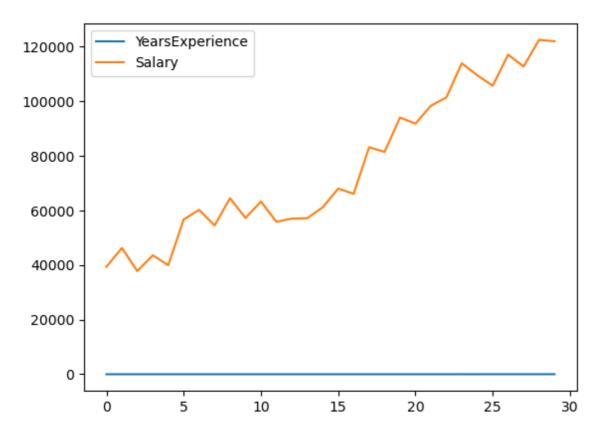


#### In [7]:

```
data.plot()
```

#### Out[7]:

<Axes: >



#### In [8]:

```
data.info()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 30 entries, 0 to 29 Data columns (total 2 columns):

# Column Non-Null Count Dtype
--- ---0 YearsExperience 30 non-null float64
1 Salary 30 non-null float64

dtypes: float64(2)

memory usage: 612.0 bytes

## In [9]:

```
x=data.iloc[:,0:1]
```

#### In [10]:

```
y=data.iloc[:,-1:]
```

#### In [11]:

```
from sklearn.model_selection import train_test_split
```

```
27/09/2023, 16:09
                            SALARY PREDICTION - LINEAR REGRESSION & DECISION TREE - Jupyter Notebook
 In [12]:
 x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=1)
 In [13]:
 print(x train.shape)
 print(x_test.shape)
 print(y_train.shape)
 print(y_test.shape)
  (21, 1)
  (9, 1)
  (21, 1)
  (9, 1)
 In [14]:
 lr=LinearRegression()
 In [15]:
 model=lr.fit(x train,y train)
 In [16]:
 prediction=lr.predict(x test)
 In [17]:
 prediction
 Out[17]:
```

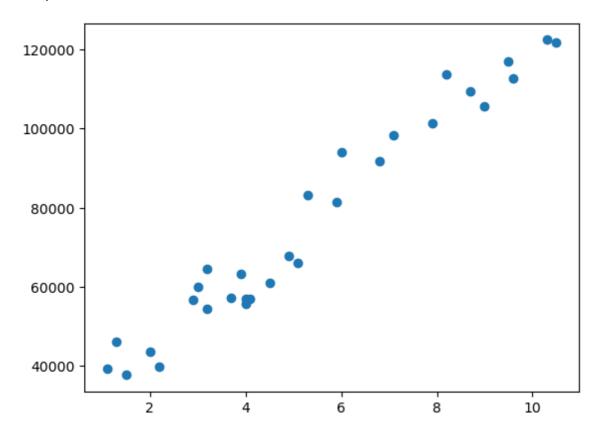
```
array([[ 74821.41578619],
       [ 91385.43626305],
       [ 61938.28874864],
       [ 81262.97930497],
       [ 67459.62890759],
       [ 88624.76618357],
       [113470.79689886],
       [ 44454.04491195],
       [106109.01002026]])
```

#### In [18]:

plt.scatter(x,y)

# Out[18]:

<matplotlib.collections.PathCollection at 0x7f5010148150>



# In [19]:

lr.coef

#### Out[19]:

array([[9202.23359825]])

# In [20]:

lr.intercept\_

## Out[20]:

array([26049.57771544])

#### In [21]:

#y=mx+c where m is slope and c= coef - constant

#### In [22]:

from sklearn import metrics

# In [23]:

from sklearn.metrics import accuracy\_score

```
In [24]:
```

```
metrics.r2_score(y_test,prediction)
```

#### Out[24]:

0.9248580247217075

# In [25]:

```
predicted_values=pd.DataFrame(prediction,columns=["Predicted"])
```

# In [26]:

# predicted\_values

#### Out[26]:

#### **Predicted**

- **0** 74821.415786
- **1** 91385.436263
- 2 61938.288749
- 3 81262.979305
- 4 67459.628908
- **5** 88624.766184
- 6 113470.796899
- **7** 44454.044912
- 8 106109.010020

#### In [29]:

# y\_test

## Out[29]:

#### Salary

- **17** 83088.0
- **21** 98273.0
- **10** 63218.0
- **19** 93940.0
- **14** 61111.0
- **20** 91738.0
- **26** 116969.0
- **3** 43525.0
- **24** 109431.0

```
In [30]:
y=9202.2*1.1+26050

In [31]:
y
Out[31]:
36172.42

In [44]:
y_pred
```

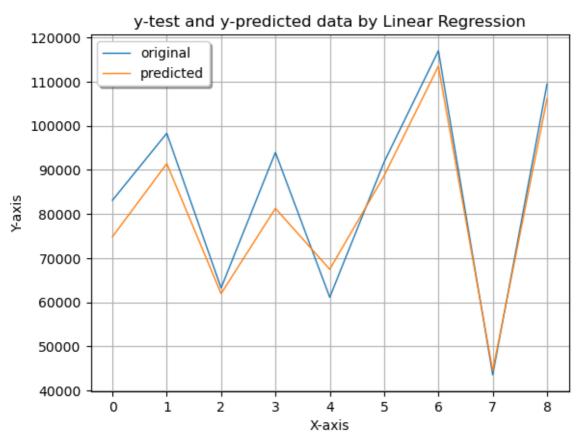
# Out[44]:

#### **Predicted**

- **0** 66029.0
- **1** 101302.0
- **2** 56375.5
- **3** 81363.0
- 4 57081.0
- **5** 81363.0
- **6** 112635.0
- 7 39891.0
- 8 105582.0

#### In [54]:

```
plt.plot(x_ax,y_test, linewidth=1, label="original")
plt.plot(x_ax,prediction,linewidth=1, label="predicted")
plt.title("y-test and y-predicted data by Linear Regression")
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.legend(loc='best',fancybox=True, shadow=True)
plt.grid(True)
plt.show()
```



#### In [47]:

```
x_ax = range(len(y_test))
```

# In [35]:

from sklearn.tree import DecisionTreeRegressor

#### In [39]:

dt=DecisionTreeRegressor()

#### In [40]:

dt.fit(x\_train,y\_train)

# Out[40]:

• DecisionTreeRegressor
DecisionTreeRegressor()

```
In [41]:
```

y\_pred=dt.predict(x\_test)

In [43]:

y\_pred=pd.DataFrame(y\_pred,columns=["Predicted"])

In [51]:

y\_pred

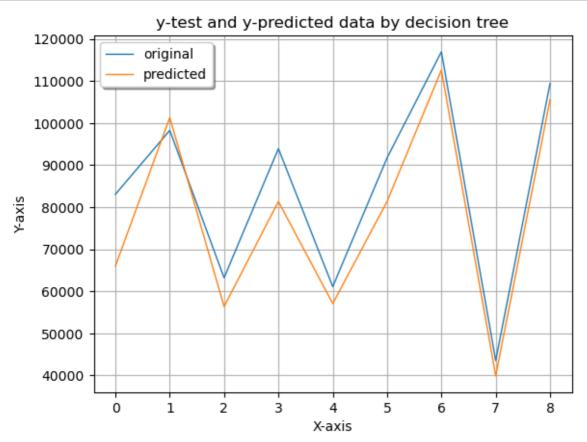
Out[51]:

## Predicted

- **0** 66029.0
- **1** 101302.0
- **2** 56375.5
- **3** 81363.0
- **4** 57081.0
- **5** 81363.0
- **6** 112635.0
- **7** 39891.0
- 8 105582.0

#### In [53]:

```
plt.plot(x_ax,y_test, linewidth=1, label="original")
plt.plot(x_ax,y_pred,linewidth=1, label="predicted")
plt.title("y-test and y-predicted data by decision tree")
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.legend(loc='best',fancybox=True, shadow=True)
plt.grid(True)
plt.show()
```



#### In [58]:

metrics.r2\_score(y\_test,y\_pred)

#### Out[58]:

0.855824415472517

#### In [ ]: