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In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
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In [2]: from sklearn.model_selection import cross_val_score
from sklearn.model_selection import KFold
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In [3]: dataset = pd.read_csv("Salary_Data.csv")
```

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In [4]: head = dataset.head() # First 5 line
print(head,"\n")
tail = dataset.tail() # Last 5 line
print(tail,"\n")
describe = dataset.describe() #summary statistics for numerical columns
print(describe,"\n")
info = dataset.info() #index, datatype and memory information
print(info,"\n")
min = dataset.min() # Returns the lowest value in each column
print(min,"\n")
median = dataset.median() # Returns the median value in each column
print(median,"\n")
max = dataset.max() # Returns the highest value in each column
print(max,"\n")

      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

      YearsExperience  Salary
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

      YearsExperience  Salary
count      30.000000    30.000000
mean         5.313333    76003.000000
std          2.837888    27414.429785
min          1.100000    37731.000000
25%          3.200000    56720.750000
50%          4.700000    65237.000000
75%          7.700000   100544.750000
max         10.500000   122391.000000

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 2 columns):
YearsExperience    30 non-null float64
Salary             30 non-null float64
dtypes: float64(2)
memory usage: 560.0 bytes
None

YearsExperience      1.1
Salary              37731.0
dtype: float64

YearsExperience      4.7
Salary              65237.0
dtype: float64

YearsExperience      10.5
Salary             122391.0
dtype: float64
```

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In [5]: X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 1].values
```

```
In [6]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 1/3, random_state = 0)
```

```
In [7]: from sklearn.linear_model import LinearRegression
```

```
In [8]: regressor = LinearRegression()
regressor.fit(X_train,y_train)
accuracies = cross_val_score(regressor, X = X_train, y = y_train, cv = 10)
print("Mean: %",accuracies.mean()*100)
print("Std: %",accuracies.std()*100)

Mean: % -259.74571918775956
std: % 646.7539113998671
```

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In [9]: y_predict = regressor.predict(X_test)
```

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In [10]: from sklearn import metrics
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In [11]: #mean absolute error
mae = metrics.mean_absolute_error(y_test,y_predict)
print("mae: ",mae)
#mean squared error
mse = metrics.mean_squared_error(y_test, y_predict)
print("mse: ",mse)
#root mean square error
rmse = np.sqrt(metrics.mean_squared_error(y_test, y_predict))
print("rmse: ",rmse)

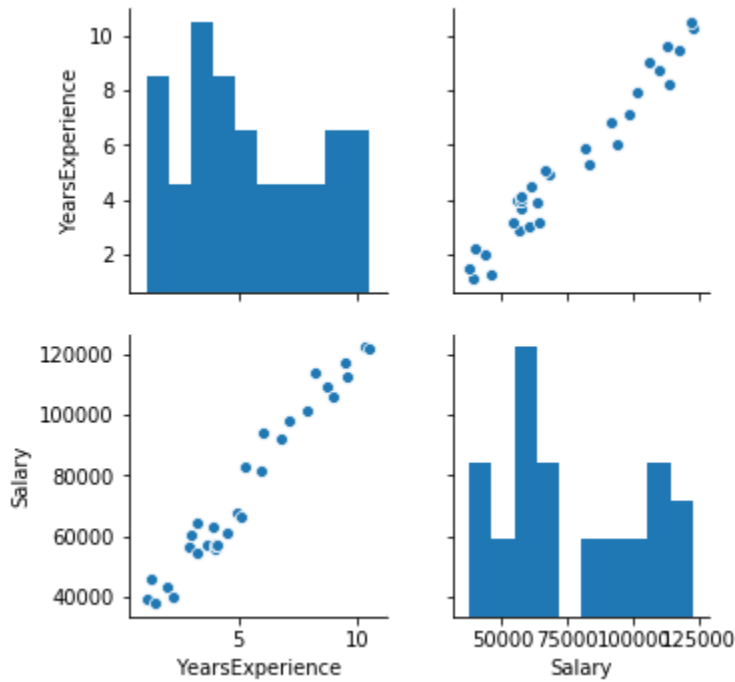
mae:  3426.4269374307123
mse:  21026037.329511296
rmse:  4585.4157204675885
```

```
In [12]: print("coef: ",regressor.coef_)

coef:  [9345.94244312]
```

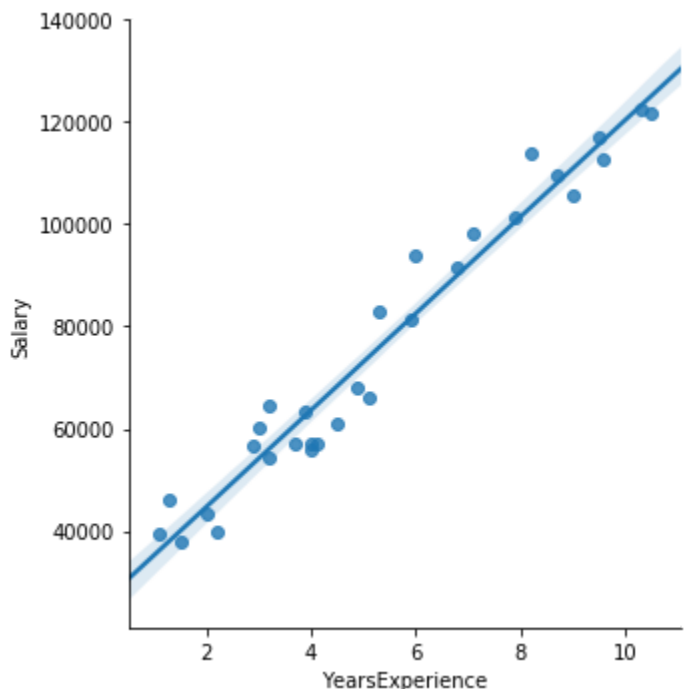
```
In [13]: sns.pairplot(dataset)
```

Out[13]: <seaborn.axisgrid.PairGrid at 0x1fe8ed17da0>



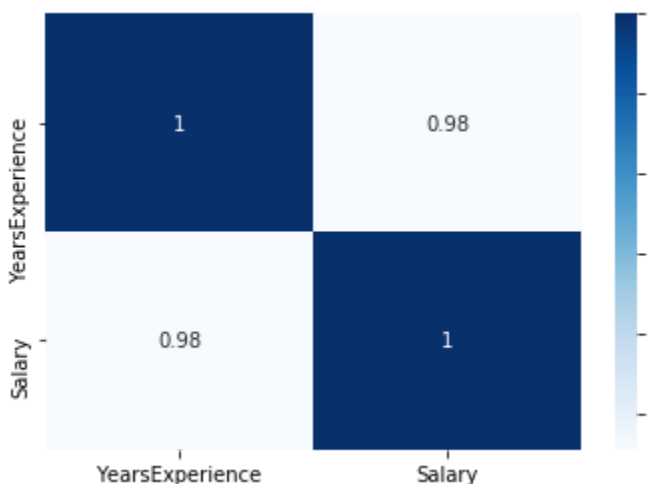
```
In [14]: sns.lmplot(x='YearsExperience',y='Salary',data=dataset)
```

Out[14]: <seaborn.axisgrid.FacetGrid at 0x1fe8f11f2e8>



```
In [15]: sns.heatmap(dataset.corr(),cmap = 'Blues', annot=True)
```

Out[15]: <matplotlib.axes._subplots.AxesSubplot at 0x1fe8f2816a0>



```
In [16]: # Visualising the Training set results
plt.scatter(X_train, y_train, color = 'red')
plt.plot(X_train, regressor.predict(X_train), color = 'blue')
plt.title('Salary vs Experience (Training set)')
plt.xlabel('Years of Experience')
plt.ylabel('Salary')
plt.show()
```



```
In [17]: # Visualising the Test set results
plt.scatter(X_test, y_test, color = 'red')
plt.plot(X_train, regressor.predict(X_train), color = 'blue')
plt.title('Salary vs Experience (Test set)')
plt.xlabel('Years of Experience')
plt.ylabel('Salary')
plt.show()
```

