

# HOUSE PRICE PREDICTION USING MACHINE LEARNING

## Importing Libraries and dataset

```
In [14]: import pandas as pd  
import matplotlib.pyplot as plt  
import numpy as np
```

```
In [15]: df=pd.read_csv("C:\\Users\\Garima Ranjan\\Downloads\\Delhi.csv")
```

## Data cleaning

```
In [16]: #Since 0 and 1 tells us that the house comes with amenities or not and 9 in the data  
df.replace(9, np.nan, inplace=True)  
df.dropna(inplace = True)  
df.isna().sum()
```

```
Out[16]: Price      0
          Area      0
          Location  0
          No. of Bedrooms  0
          Resale     0
          MaintenanceStaff  0
          Gymnasium  0
          SwimmingPool  0
          LandscapedGardens  0
          JoggingTrack  0
          RainWaterHarvesting  0
          IndoorGames  0
          ShoppingMall  0
          Intercom    0
          SportsFacility  0
          ATM         0
          ClubHouse   0
          School      0
          24X7Security  0
          PowerBackup  0
          CarParking  0
          StaffQuarter  0
          Cafeteria   0
          MultipurposeRoom  0
          Hospital    0
          WashingMachine  0
          Gasconnection  0
          AC          0
          Wifi        0
          Children'splayarea  0
          LiftAvailable  0
          BED         0
          VaastuCompliant  0
          Microwave   0
          GolfCourse  0
          TV          0
          DiningTable  0
          Sofa        0
          Wardrobe    0
          Refrigerator  0
          dtype: int64
```

```
In [17]: df.shape
```

```
Out[17]: (2002, 40)
```

## Data Processing

```
In [18]: #finding number of objects,integers and floats
obj=(df.dtypes=='object')
object_cols=list(obj[obj].index)
print("Categorical variables:", len(object_cols))

int_=(df.dtypes=='int64')
int_cols=list(int_[int_].index)
print("Integer variables:",len(int_cols))

fl_=(df.dtypes=='float64')
fl_cols=list(fl_[fl_].index)
print("Float variables:",len(fl_cols))
```

Categorical variables: 1  
 Integer variables: 4  
 Float variables: 35

In [19]: `df.describe()`

Out[19]:

	Price	Area	No. of Bedrooms	Resale	MaintenanceStaff	Gymnasium	Swim
<b>count</b>	2.002000e+03	2002.000000	2002.000000	2002.000000	2002.000000	2002.000000	20
<b>mean</b>	1.029472e+07	1200.141359	2.688811	0.700799	0.071928	0.193806	
<b>std</b>	9.119848e+06	614.340764	0.729588	0.458022	0.258433	0.395378	
<b>min</b>	2.000000e+06	200.000000	1.000000	0.000000	0.000000	0.000000	
<b>25%</b>	3.600000e+06	720.000000	2.000000	0.000000	0.000000	0.000000	
<b>50%</b>	7.000000e+06	1000.000000	3.000000	1.000000	0.000000	0.000000	
<b>75%</b>	1.550000e+07	1700.000000	3.000000	1.000000	0.000000	0.000000	
<b>max</b>	1.625000e+08	6400.000000	5.000000	1.000000	1.000000	1.000000	

8 rows × 39 columns

## Encoding Categorical data

In [20]: `from sklearn.preprocessing import OneHotEncoder`

```
s = (df.dtypes == 'object')
object_cols = list(s[s].index)
print("Categorical variables:")
print(object_cols)
print('No. of. categorical features: ',
      len(object_cols))
```

Categorical variables:  
 ['Location']  
 No. of. categorical features: 1

In [21]: `OH_encoder = OneHotEncoder(sparse=False)`  
`OH_cols = pd.DataFrame(OH_encoder.fit_transform(df[object_cols]))`  
`OH_cols.index = df.index`  
`OH_cols.columns = OH_encoder.get_feature_names_out()`  
`df_final = df.drop(object_cols, axis=1)`  
`df_final = pd.concat([df_final, OH_cols], axis=1)`

In [22]: `df_final`

Out[22]:

	Price	Area	No. of Bedrooms	Resale	MaintenanceStaff	Gymnasium	SwimmingPool	Landscaping
0	10500000	1200	2	1	0.0	1.0	0.0	
1	6000000	1000	3	0	0.0	0.0	0.0	
2	15000000	1350	2	1	0.0	0.0	0.0	
3	2500000	435	2	0	0.0	0.0	0.0	
4	5800000	900	3	0	0.0	0.0	0.0	
...	...	...	...	...	...	...	...	...
1997	9000000	1200	2	1	0.0	1.0	1.0	
1998	12000000	1350	3	1	0.0	0.0	0.0	
1999	16000000	1963	4	1	0.0	0.0	0.0	
2000	11600000	1050	2	1	0.0	1.0	0.0	
2001	13600000	1450	3	1	0.0	0.0	0.0	

2002 rows × 142 columns

## Splitting Dataset into training and testing

```
In [23]: from sklearn.metrics import mean_absolute_error
from sklearn.model_selection import train_test_split

X = np.array(df_final.drop(['Price'], axis=1))
Y = np.array(df_final['Price'])
Y = Y.reshape(len(Y),1)
# Split the training set into
# training and validation set
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, train_size=0.8, test_size=0.2)
```

## Using Support Vector Regression(SVR) Model

```
In [24]: from sklearn.metrics import mean_absolute_percentage_error

from sklearn.preprocessing import StandardScaler
sc_X = StandardScaler()
sc_y = StandardScaler()
X_train = sc_X.fit_transform(X_train)
Y_train = sc_y.fit_transform(Y_train)
```

```
In [25]: from sklearn.svm import SVR
regressor = SVR(kernel = 'rbf')
regressor.fit(X_train, Y_train)
```

```
C:\Users\Garima Ranjan\anaconda3\lib\site-packages\sklearn\utils\validation.py:99:
DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
y = column_or_1d(y, warn=True)
```

Out[25]: SVR()

```
In [26]: #predicting the test set results  
Y_pred= regressor.predict(X_test)  
print(mean_absolute_percentage_error(Y_test, Y_pred))  
  
0.9999999540899248
```

```
In [27]: Y_pred = sc_y.inverse_transform(regressor.predict(sc_X.transform(X_test)).reshape(  
np.set_printoptions(precision=2)  
print(np.concatenate((Y_pred.reshape(len(Y_pred),1), Y_test.reshape(len(Y_test),1)
```

```
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[16442522.49 18000000. ]
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```

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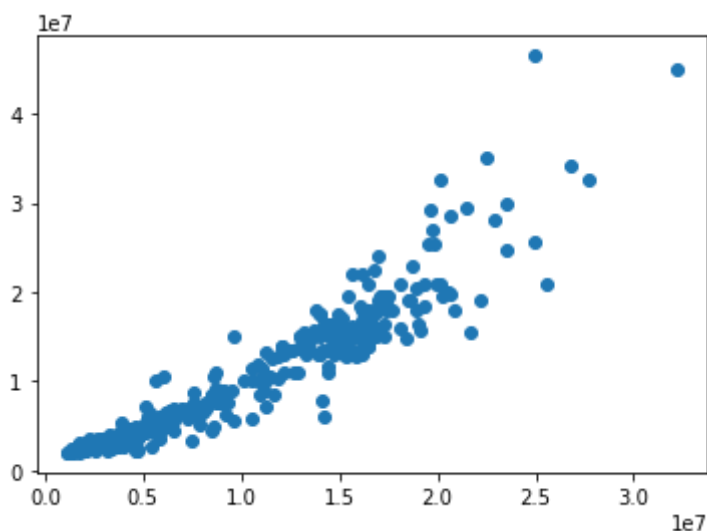
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```
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```
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```

In [28]: `plt.scatter(Y_pred, Y_test)`

Out[28]: <matplotlib.collections.PathCollection at 0x25c7e29dfa0>



## Using Random Forest Model

In [61]: `#resetting values of y_pred and x_pred`  
`X_train, X_test, Y_train, Y_test = train_test_split(X, Y, train_size=0.8, test_size=0.2)`

In [62]: `from sklearn.metrics import mean_absolute_percentage_error`  
`from sklearn.ensemble import RandomForestRegressor`  
`regressor2 = RandomForestRegressor(n_estimators = 500, random_state = 0)`  
`regressor2.fit(X_train, Y_train)`

C:\Users\Garima Ranjan\AppData\Local\Temp\ipykernel\_20444\1903883464.py:4: DataCon  
versionWarning: A column-vector y was passed when a 1d array was expected. Please  
change the shape of y to (n\_samples,), for example using ravel().  
`regressor2.fit(X_train, Y_train)`

Out[62]: RandomForestRegressor(n\_estimators=500, random\_state=0)

In [63]: `Y_pred = regressor2.predict(X_test)`  
`print(mean_absolute_percentage_error(Y_test, Y_pred))`

0.11664613066282309

```
In [64]: np.set_printoptions(precision=2)
print(np.concatenate((Y_pred.reshape(len(Y_pred),1), Y_test.reshape(len(Y_test),1))
```

```
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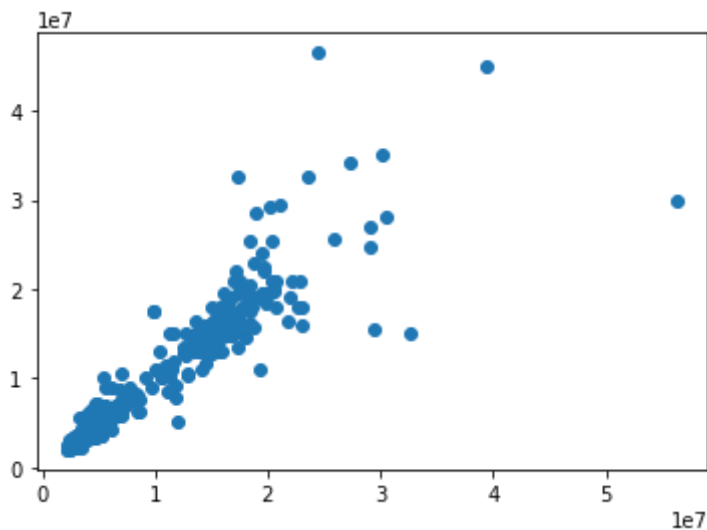
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```
In [46]: plt.scatter(Y_pred,Y_test)
```

```
Out[46]: <matplotlib.collections.PathCollection at 0x25c0096a190>
```



Random forest model is best with error of just 0.115

## Challenges faced

```
In [48]: #Finding dataset
          #Number of data in dataset
          #Finding the right model
          #increasing accuracy of Random Forest
```