KNN- Classifier

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We will do Knn classifier on our same housing market data.

Importing all the needed modules:

```
import pandas as pd
import numpy as np
from sklearn import preprocessing, neighbors, model_selection, metrics
import matplotlib.pyplot as plt
import seaborn as sb
```

Reading Data

```
#Importing our training and testing house prices data set.
df=pd.read_excel("/Users/devmarwah/Downloads/BA-Predict-2.xlsx")
```

```
#Printing head of training dataset:
df.head()
```

	LotArea	OverallQual	YearBuilt	YearRemodAdd	BsmtFinSF1	FullBath	HalfBath	BedroomAbvGr	TotRmsAbvGrd	Fireplace
0	7340	4	1971	1971	322	1	0	2	4	
1	8712	5	1957	2000	860	1	0	2	5	
2	7875	7	2003	2003	0	2	1	3	8	
3	14859	7	2006	2006	0	2	0	3	7	
4	6173	5	1967	1967	599	1	0	3	6	

Data cleaning:

Checking for any missing values in our dataset:

```
df.isna().sum()
```

```
LotArea         0
OverallQual     0
YearBuilt       0
YearRemodAdd    0
BsmtFinSF1      0
FullBath        0
HalfBath        0
BedroomAbvGr    0
TotRmsAbvGrd    0
Fireplaces      0
GarageArea      0
YrSold          0
SalePrice       0
dtype: int64
```

Hence, there are no missing values in our dataset.

Now, checking for duplicates in the dataset.

```
df.duplicated().sum()
```

```
0
```

Hence, there are no dubplicate values in our dataset.

Data prepration:

```
df.head()
```

	LotArea	OverallQual	YearBuilt	YearRemodAdd	BsmtFinSF1	FullBath	HalfBath	BedroomAbvGr	TotRmsAbvGrd	Fireplace
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3	14859	7	2006	2006	0	2	0	3	7	
4	6173	5	1967	1967	599	1	0	3	6	

In our dataset, following variables are categorical: "OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"
Hence, we need to convert them to category dtype.

```
df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBat
#Printing dtypes of our data to verify:
df.dtypes
```

```
LotArea           int64
OverallQual    category
```

```
YearBuilt          int64
YearRemodAdd       int64
BsmtFinSF1         int64
FullBath        category
HalfBath        category
BedroomAbvGr    category
TotRmsAbvGrd    category
Fireplaces      category
GarageArea         int64
YrSold          category
SalePrice          int64
dtype: object
```

Hence, all categorical variables have been converted to category dtype now.

```
#Variable OverallQual just rates overall quality and is not necessary in our analysis so we will drop it
df.drop('OverallQual',axis=1,inplace=True)
```

```
#Printing count of distinct categories in variables to find variables which should be converted as dummy variables:
print(
df['FullBath'].unique(),
df['HalfBath'].unique(),
df['BedroomAbvGr'].unique(),
df['TotRmsAbvGrd'].unique(),
df['Fireplaces'].unique(),
df['YrSold'].unique())
```

```
[1, 2, 0]
Categories (3, int64): [0, 1, 2] [0, 1, 2]
Categories (3, int64): [0, 1, 2] [2, 3, 5, 4, 1]
Categories (5, int64): [1, 2, 3, 4, 5] [4, 5, 8, 7, 6, 12, 9, 10, 11]
Categories (9, int64): [4, 5, 6, 7, ..., 9, 10, 11, 12] [0, 1, 2]
Categories (3, int64): [0, 1, 2] [2007, 2009, 2006, 2010, 2008]
Categories (5, int64): [2006, 2007, 2008, 2009, 2010]
```

```
#Here totrmsabvgrd has 9 categories and can make our model complex hence we are dropping it:
df.drop('TotRmsAbvGrd',axis=1,inplace=True)
```

Turning other categorical variables into dummy variables as all of them have more than two categories;

```
df=pd.get_dummies(df,columns=["FullBath","HalfBath","BedroomAbvGr","Fireplaces","YrSold"]).astype('int')
```

```
#Printing head() of dataset to verify dummy variables:
columns=df.columns[[0,1,2,3,4,5]]
columns
```

```
    Index(['LotArea', 'YearBuilt', 'YearRemodAdd', 'BsmtFinSF1', 'GarageArea',
           'SalePrice'],
          dtype='object')
```

Normalizing our data:

```
df_norm=pd.DataFrame(preprocessing.StandardScaler().fit_transform(df[['LotArea','YearBuilt','YearRemodAdd','BsmtFinSF1',
df_norm=pd.DataFrame(df_norm)
df_norm.columns=columns
#Replacing normalized columns in our main dataframe
df[['LotArea','YearBuilt','YearRemodAdd','BsmtFinSF1','GarageArea','SalePrice']]=df_norm[['LotArea','YearBuilt','YearRem
df.head()
```

	LotArea	YearBuilt	YearRemodAdd	BsmtFinSF1	GarageArea	SalePrice	FullBath_0	FullBath_1	FullBath_2	HalfBath_0
0	-0.546080	-0.109679	-0.668838	-0.260487	0.994166	-1.028489	0	1	0	1
1	-0.230414	-0.598001	0.714417	1.085306	1.337347	-0.321871	0	1	0	1
2	-0.422989	1.006486	0.857512	-1.065961	-0.392858	0.121820	0	0	1	0
3	1.183869	1.111127	1.000608	-1.065961	1.022764	1.107798	0	0	1	1
4	-0.814580	-0.249199	-0.859632	0.432421	-0.893330	-0.773778	0	1	0	1

5 rows × 25 columns

Knn classification is applied on classes. Hence, we will convert Saleprice into three categories.

```
#Coverting SalePrice into 'High' and 'Low' class:
df['Price']=pd.DataFrame(np.where(df['SalePrice']>=df['SalePrice'].mean(),'High','Low'))
df.head()
```

	LotArea	YearBuilt	YearRemodAdd	BsmtFinSF1	GarageArea	SalePrice	FullBath_0	FullBath_1	FullBath_2	HalfBath_0
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4	-0.814580	-0.249199	-0.859632	0.432421	-0.893330	-0.773778	0	1	0	1

5 rows × 26 columns

Making training and testing data sets:

```
#Dropping SalePrice first
df.drop('SalePrice',axis=1,inplace=True)


#Making training and testing dataset now:
x=np.array(df.drop('Price',axis=1))
y=np.array(df['Price'])
t=model_selection.train_test_split
x_train, x_test, y_train, y_test = t(x,y,test_size=0.2,random_state=2)
```

Model Construction

Our data is finally prepared and we can apply knn-classifier model :

```
clf=neighbors.KNeighborsClassifier(n_neighbors=2)
clf.fit(x_train,y_train)
```

```
▾ KNeighborsClassifier
KNeighborsClassifier(n_neighbors=2)
```

Printing accuracy of our model:

```
print(clf.score(x_test,y_test))
```

```
0.8333333333333334
```

Having a look at predictions made:

```
p=clf.predict(x_test)
```

Making confusion matrix of our model:

```
cm=metrics.confusion_matrix
c=cm(y_test,p)
plt.figure()
sb.heatmap(c,annot=True,cmap="Blues",xticklabels=['Low','High'],yticklabels=['Low','High'])
plt.xlabel('Actual values')
plt.ylabel('Predicted Values')
```

```
Text(50.722222222222214, 0.5, 'Predicted Values')
```