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
```

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	Fireplaces	GarageArea	YrSold	SalePrice
0	7340	4	1971	1971	322	1	0	2	4	0	684	2007	110000
1	8712	5	1957	2000	860	1	0	2	5	0	756	2009	153000
2	7875	7	2003	2003	0	2	1	3	8	1	393	2006	180000
3	14859	7	2006	2006	0	2	0	3	7	1	690	2006	240000
4	6173	5	1967	1967	599	1	0	3	6	0	288	2007	125500

Data cleaning:

Checking for any missing values in our dataset:

df.isna().sum()

LotArea 0 OverallQual 0 YearBuilt YearRemodAdd BsmtFinSF1 0 FullBath HalfBath 0 BedroomAbvGr 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
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	

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

```
df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","Fireplaces","YrSold"]]=df[["OverallQual","FullBath","HalfBath","BedroomAbvGr","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRmsAbvGrd","TotRm
```

```
LotArea
                   int64
OverallOual
                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['HalfBath'].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 (5, int64): [2006, 2007, 2008, 2009, 2010]

Categories (3, int64): [0, 1, 2] [2007, 2009, 2006, 2010, 2008]

#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;

Normalizing our data:

```
df_norm=pd.DataFrame(preprocessing.StandardScaler().fit_transform(df[['LotArea', 'YearBuilt', 'YearRemodAdd', 'BsmtFinSF1', 'GarageArea', 'SalePrice']]))
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', 'YearRemodAdd', 'BsmtFinSF1', 'GarageArea'
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
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 × 26 columns

Making training and testing data sets: