

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import r2_score
from sklearn.metrics import accuracy_score
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: house_price=pd.read_csv("train.csv")
```

```
In [3]: house_price.head()
```

	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	...	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	Sale
0	1	60	RL	65.0	8450	Pave	NaN	Reg	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	2008	WD	
1	2	20	RL	80.0	9600	Pave	NaN	Reg	Lvl	AllPub	...	0	NaN	NaN	NaN	0	5	2007	WD	
2	3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	9	2008	WD	
3	4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	2006	WD	
4	5	60	RL	84.0	14260	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	12	2008	WD	

5 rows × 81 columns

```
In [4]: # Data pre-processing
house_price.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 81 columns):
#   Column              Non-Null Count  Dtype
---  --
0   Id                   1460 non-null   int64
1   MSSubClass           1460 non-null   int64
2   MSZoning              1460 non-null   object
3   LotFrontage          1201 non-null   float64
4   LotArea              1460 non-null   int64
5   Street               1460 non-null   object
6   Alley                91 non-null     object
7   LotShape             1460 non-null   object
8   LandContour          1460 non-null   object
9   Utilities             1460 non-null   object
10  LotConfig            1460 non-null   object
11  Landslope            1460 non-null   object
12  Neighborhood         1460 non-null   object
13  Condition1           1460 non-null   object
14  Condition2           1460 non-null   object
15  BldgType             1460 non-null   object
16  HouseStyle           1460 non-null   object
17  OverallQual           1460 non-null   int64
18  OverallCond           1460 non-null   int64
19  YearBuilt             1460 non-null   int64
20  YearRemodAdd          1460 non-null   int64
21  RoofStyle             1460 non-null   object
22  RoofMatl             1460 non-null   object
23  Exterior1st           1460 non-null   object
24  Exterior2nd           1460 non-null   object
25  MasVnrType           1452 non-null   object
26  MasVnrArea            1452 non-null   float64
27  ExterQual             1460 non-null   object
28  ExterCond             1460 non-null   object
29  Foundation            1460 non-null   object
30  BsmtQual              1423 non-null   object
31  BsmtCond              1423 non-null   object
32  BsmtExposure          1422 non-null   object
33  BsmtFinType1          1423 non-null   object
34  BsmtFinSF1            1460 non-null   int64
35  BsmtFinType2          1422 non-null   object
36  BsmtFinSF2            1460 non-null   int64
37  BsmtUnfSF             1460 non-null   int64
38  TotalBsmtSF           1460 non-null   int64
39  Heating              1460 non-null   object
40  HeatingQC             1460 non-null   object
41  CentralAir            1460 non-null   object
42  Electrical            1459 non-null   object
43  1stFlrSF              1460 non-null   int64
44  2ndFlrSF              1460 non-null   int64
45  LowQualFinSF          1460 non-null   int64
46  GrLivArea             1460 non-null   int64
47  BsmtFullBath          1460 non-null   int64
48  BsmtHalfBath          1460 non-null   int64
49  FullBath              1460 non-null   int64
50  HalfBath              1460 non-null   int64
51  BedroomAbvGr          1460 non-null   int64
52  KitchenAbvGr          1460 non-null   int64
53  KitchenQual           1460 non-null   object
54  TotRmsAbvGrd          1460 non-null   int64
55  Functional            1460 non-null   object
56  Fireplaces            1460 non-null   int64
57  FireplaceQu           770 non-null   object
58  GarageType            1379 non-null   object
59  GarageYrBlt           1379 non-null   float64
60  GarageFinish          1379 non-null   object
61  GarageCars            1460 non-null   int64
62  GarageArea            1460 non-null   int64
63  GarageQual            1379 non-null   object
64  GarageCond            1379 non-null   object
65  PavedDrive            1460 non-null   object
66  WoodDeckSF            1460 non-null   int64
67  OpenPorchSF           1460 non-null   int64
68  EnclosedPorch         1460 non-null   int64
69  3SsnPorch             1460 non-null   int64
70  ScreenPorch           1460 non-null   int64
71  PoolArea              1460 non-null   int64
72  PoolQC                7 non-null     object
73  Fence                 281 non-null   object
74  MiscFeature           54 non-null   object
75  MiscVal               1460 non-null   int64
76  MoSold                1460 non-null   int64
77  YrSold                1460 non-null   int64
78  SaleType              1460 non-null   object
79  SaleCondition         1460 non-null   object
80  SalePrice             1460 non-null   int64
dtypes: float64(3), int64(35), object(43)
memory usage: 924.0+ KB
```

```
In [5]: house_price.describe()
```

	Id	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemodAdd	MasVnrArea	BsmtFinSF1	...	WoodDeckSF	OpenPorchSF	Enclose
count	1460.000000	1460.000000	1201.000000	1460.000000	1460.000000	1460.000000	1460.000000	1460.000000	1452.000000	1460.000000	...	1460.000000	1460.000000	1460.
mean	730.500000	56.897260	70.049958	10516.820882	6.099315	5.575342	1971.267808	1984.865753	100.685262	443.639726	...	94.244521	46.660274	21.
std	421.610009	42.300571	24.284752	9981.264932	1.382997	1.112759	30.202904	20.445407	181.066207	456.098091	...	125.338794	66.256028	61.
min	1.000000	20.000000	21.000000	1300.000000	1.000000	1.000000	1872.000000	1950.000000	0.000000	0.000000	...	0.000000	0.000000	0.
25%	365.750000	20.000000	59.000000	7553.500000	5.000000	5.000000	1954.000000	1967.000000	0.000000	0.000000	...	0.000000	0.000000	0.
50%	730.500000	50.000000	69.000000	9478.500000	6.000000	5.000000	1973.000000	1994.000000	0.000000	383.500000	...	0.000000	25.000000	0.
75%	1095.250000	70.000000	80.000000	11601.500000	7.000000	6.000000	2000.000000	2004.000000	166.000000	712.250000	...	168.000000	68.000000	0.
max	1460.000000	190.000000	313.000000	215245.000000	10.000000	9.000000	2010.000000	2010.000000	1600.000000	5644.000000	...	857.000000	547.000000	552.

8 rows × 38 columns

```
In [6]: house_price.columns
```

```
Out[6]: Index(['Id', 'MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
        'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig',
        'Landslope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType',
        'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd',
        'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType',
        'MasVnrArea', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual',
        'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1',
        'BsmtFinType2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating',
        'HeatingQC', 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF',
        'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath',
        'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual',
        'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType',
        'GarageYrBlt', 'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual',
        'GarageCond', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF',
        'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'PoolQC',
        'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleType',
        'SaleCondition', 'SalePrice'],
        dtype='object')
```

```
In [7]: house_price.shape
```

```
Out[7]: (1460, 81)
```

```
In [8]: #how many null vallues is columns
house_price.isnull().sum()
```

Id	0
MSSubClass	0
MSZoning	0
LotFrontage	259
LotArea	0
...	0
MoSold	0
YrSold	0
SaleType	0
SaleCondition	0
SalePrice	0
Length:	81, dtype: int64

```
In [9]: sns.heatmap(house_price.isnull(),yticklabels=False,cbar=False)
```

```
Out[9]: <AxesSubplot:~>
```



```
In [10]: #fill all missing values
house_price['LotFrontage']=house_price['LotFrontage'].fillna(house_price['LotFrontage'].mean())
```

```
In [11]: house_price.drop(['Alley'],axis=1,inplace=True)
```

```
In [12]: house_price['BsmtQual']=house_price['BsmtQual'].fillna(house_price['BsmtQual'].mode()[0])
house_price['BsmtQual']=house_price['BsmtQual'].fillna(house_price['BsmtQual'].mode()[0])
```

```
In [13]: house_price['FireplaceQu']=house_price['FireplaceQu'].fillna(house_price['FireplaceQu'].mode()[0])
house_price['GarageType']=house_price['GarageType'].fillna(house_price['GarageType'].mode()[0])
```

```
In [14]: house_price.drop(['GarageYrBlt'],axis=1,inplace=True)
```

```
In [15]: house_price['GarageFinish']=house_price['GarageFinish'].fillna(house_price['GarageFinish'].mode()[0])
house_price['GarageQual']=house_price['GarageQual'].fillna(house_price['GarageQual'].mode()[0])
house_price['GarageCond']=house_price['GarageCond'].fillna(house_price['GarageCond'].mode()[0])
```

```
In [16]: house_price.drop(['PoolQC','Fence','MiscFeature'],axis=1,inplace=True)
```

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

```
Out[17]: (1460, 76)
```

```
In [18]: #Remove all null values
house_price.drop(['Id'],axis=1,inplace=True)
```

```
In [19]: house_price.isnull().sum()
```

MSSubClass	0
MSZoning	0
LotFrontage	0
LotArea	0
Street	0
...	0
MoSold	0
YrSold	0
SaleType	0
SaleCondition	0
SalePrice	0
Length:	75, dtype: int64

```
In [71]: house_price['MasVnrType']=house_price['MasVnrType'].fillna(house_price['MasVnrType'].mode()[0])
house_price['MasVnrArea']=house_price['MasVnrArea'].fillna(house_price['MasVnrArea'].mode()[0])
```

```
In [72]: sns.heatmap(house_price.isnull(),yticklabels=False,chr=False,cmap='coolwarm')
```

```
Out[72]: <AxesSubplot:~>
```



```
In [73]: house_price['BsmtExposure']=house_price['BsmtExposure'].fillna(house_price['BsmtExposure'].mode()[0])
```

```
In [74]: sns.heatmap(house_price.isnull(),yticklabels=False,chr=False,cmap='YlOrBu')
```

```
Out[74]: <AxesSubplot:~>
```



```
In [24]: house_price['BsmtFinType2']=house_price['BsmtFinType2'].fillna(house_price['BsmtFinType2'].mode()[0])
```

```
In [25]: house_price.dropna(inplace=True)
```

```
In [26]: house_price.shape
```

```
Out[26]: (1422, 75)
```

```
In [27]: ## Handling categorical feature
columns=['MSZoning', 'LotFrontage', 'LotArea', 'Street', 'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'Landslope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType', 'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType', 'MasVnrArea', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', 'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'HeatingQC', 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType', 'GarageYrBlt', 'GarageFinish', 'GarageCars', 'GarageQual', 'GarageCond', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'PoolQC', 'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleType', 'SaleCondition', 'SalePrice']
```

```
In [28]: len(columns)
```

```
Out[28]: 79
```

```
In [29]: def category_multiple(multicolumns):
df_final=df
for fields in multicolumns:
    print(fields)
    df=pd.get_dummies(df_final,drop_first=True)
    df=df.drop([fields],axis=1,inplace=True)
    if i==0:
        df_final=df.copy()
    else:
        df_final=pd.concat([df_final,df],axis=1)
        df_final=pd.concat([df_final,df_final],axis=1)
    return df_final
```

```
In [30]: main_df=house_price.copy()
```

```
In [31]: # Combine the test data
test_data=pd.read_csv('formulate.csv',sep='\\s',\\s+',header=0, encoding='ascii', engine='python')
```

```
In [32]: test_data.shape
```

```
Out[32]: (93, 76)
```

```
In [33]: test_data.head()
```

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	LotConfig	...	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	PoolArea	MiscV
0	160	FV	39.0	3215	Pave	Pave	Reg	Lvl	AllPub	Inside	...	111	0	0	0	0	0
1	160	FV	30.0	9515	Pave	Pave	Reg	Lvl	AllPub	Inside	...	172	0	0	0	0	0
2	160	FV	24.0	2544	Pave	Pave	Reg	Lvl	AllPub	Inside	...	172	0	0	0	0	0
3	160	FV	24.0	2544	Pave	Pave	Reg	Lvl	AllPub	Inside	...	166	0	0	0	0	0
4	20	FV	57.0	12853	Pave	Pave	IR1	Lvl	AllPub	Inside	...	192	0	224	0	0	0

5 rows × 76 columns

```
In [34]: final_data=pd.concat([house_price,test_data],axis=0)
```

```
In [35]: print(final_data.columns)
```

```
Index(['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street',
        'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'Landslope',
        'Neighborhood', 'Condition1', 'Condition2', 'BldgType', 'HouseStyle',
        'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'RoofStyle',
        'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType', 'MasVnrArea',
        'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', 'BsmtCond',
        'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2',
        'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'HeatingQC', 'CentralAir',
        'Electrical', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF', 'GrLivArea',
        'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath',
        'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd',
        'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType', 'GarageFinish',
        'GarageCars', 'GarageQual', 'GarageCond', 'PavedDrive', 'WoodDeckSF',
        'OpenPorchSF', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch',
        'PoolArea', 'PoolQC', 'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleType',
        'SaleCondition', 'SalePrice'],
        dtype='object')
```

```
In [36]: final_data.shape
```

```
Out[36]: (1515, 77)
```

```
In [37]: data_train=final_data.iloc[:1490,:]
data_test=final_data.iloc[1490:]
```

```
In [38]: data_test.drop(['SalePrice'],axis=1,inplace=True)
```

```
In [39]: data_test.shape
```

```
Out[39]: (115, 76)
```

```
In [58]: ## Predction of data set using algorithm
x=data_train[['SalePrice']]
y=data_train[['SalePrice']]
```

```
In [59]: X_train,X_test,Y_train,Y_test = train_test_split(x,y,test_size=0.1,random_state=1)
```

```
In [60]: print(X_train)
print(X_test)
```

	SalePrice
378	394432.0
638	85800.0
1425	142800.0
1358	177500.0
174	184000.0
...	...
732	222500.0
935	235500.0
1127	250900.0
241	110500.0
1092	135500.0

[1260 rows x 1 columns]

	SalePrice
562	108000.0
773	114500.0
1318	175000.0
1159	185000.0
1094	129000.0
...	...
594	110000.0
1407	112000.0
471	190000.0
916	35311.0
1161	224000.0

[140 rows x 1 columns]

```
In [43]: # Training the model
le =LinearRegression()
#classifier=LinearRegression()
```

```
In [44]: #Training model with fit line
le.fit(X_train,y_train)
```

```
Out[44]: LinearRegression
LinearRegression()
```

```
In [53]: le.intercept_
```

```
Out[53]: 193578.75621028928
```

```
In [54]: le.coef_
```

```
Out[54]: array([-180.87121729])
```

```
In [61]: ##Accuracy of testing data
le.score(X_test,y_test)
y_prediction=le.predict(X_test)
```

```
In [66]: print(r2_score(y_prediction, X_test))
```

-4.421377338775149

```
In [50]: import pandas as pd
data=pd.read_csv('train.csv')
data
```

data = pd.read_csv('train.csv')																				
	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	...	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	
0	1	60	RL	65.0	8450	Pave	NaN	Reg	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	2008	WD	
1	2	20	RL	80.0	9600	Pave	NaN	Reg	Lvl	AllPub	...	0	NaN	NaN	NaN	0	5	2007	WD	
2	3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	9	2008	WD	
3	4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	2	2006	WD	
4	5	60	RL	84.0	14260	Pave	NaN	IR1	Lvl	AllPub	...	0	NaN	NaN	NaN	0	12	2008	WD	

1455 1456 20 60 RL 65.0 8450 8450 Pave NaN Reg Lvl AllPub Inside ... 0 NaN NaN NaN NaN 0 8 2007 WD

1456 1457 20 60 RL 68.0 11250 11250 Pave NaN Reg Lvl AllPub Inside ... 0 NaN MmPrv NaN NaN 0 2 2010 WD

1457 1458 70 70 RL 66.0 9042 9042 Pave NaN Reg Lvl AllPub Inside ... 0 NaN GdPrv Shed 2500 5 2010 WD