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from sklearn.linear_model import LinearRegression
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler, PolynomialFeatures
from sklearn.linear_model import Ridge

file_name='kc_house_data.csv'
df=pd.read_csv(file_name)

print('-----1-----')
print(df.dtypes)
print(df.describe())
print('-----2-----')
# df.drop(["id","Unnamed: 0"], axis = 1,inplace = True)
print(df.describe())
print('-----3-----')
print(df['floors'].value_counts().to_frame())
print('-----4-----')
sns.boxplot(x="floors", y="price", data=df)
plt.show()
print('-----5-----')
sns.regplot(x="sqft_above", y="price", data=df)
plt.ylim(0,)
plt.show()
print('-----6-----')
X = df[['sqft_living']]
Y = df['price']
lm = LinearRegression()
lm
lm.fit(X,Y)
print(lm.score(X, Y))
print('-----7-----')

features = ["floors", "waterfront","lat" ,"bedrooms" ,"sqft_basement" ,"view"
,"bathrooms","sqft_living15","sqft_above","grade","sqft_living"]
Multi_X = df[features]
lm1 = LinearRegression()
lm1
lm1.fit(Multi_X,Y)
print(lm1.score(Multi_X, Y))
print('-----8-----')
Input=[('scale',StandardScaler()),('polynomial',
PolynomialFeatures(include_bias=False)),('model',LinearRegression())]
pipe=Pipeline(Input)
pipe.fit(Multi_X,Y)
print(pipe.score(Multi_X,Y))

print('-----9-----')

from sklearn.model_selection import cross_val_score
from sklearn.model_selection import train_test_split

features = ["floors", "waterfront","lat" ,"bedrooms" ,"sqft_basement" ,"view"
,"bathrooms","sqft_living15","sqft_above","grade","sqft_living"]
X = df[features ]
Y = df['price']

x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.15,

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random_state=1)

print("number of test samples :", x_test.shape[0])
print("number of training samples:",x_train.shape[0])

Ridge_obj = Ridge(alpha=0.1)
Ridge_obj.fit(x_train,y_train)
Ridge_obj.score(x_test, y_test)

print('-----10-----')

Poly_obj=PolynomialFeatures(degree=2)
x_train_pr=Poly_obj.fit_transform(x_train)
x_test_pr=Poly_obj.fit_transform(x_test)
Ridge_obj = Ridge(alpha=0.1)
Ridge_obj.fit(x_train_pr,y_train)
print(Ridge_obj.score(x_test_pr, y_test))

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-----1-----
id                int64
date              object
price             float64
bedrooms          int64
bathrooms         float64
sqft_living       int64
sqft_lot          int64
floors            float64
waterfront        int64
view              int64
condition         int64
grade             int64
sqft_above        int64
sqft_basement     int64
yr_built          int64
yr_renovated      int64
zipcode           int64
lat               float64
long              float64
sqft_living15     int64
sqft_lot15        int64
dtype: object

      id      price  ...  sqft_living15  sqft_lot15
count  2.161300e+04  2.161300e+04  ...    21613.000000    21613.000000
mean   4.580302e+09  5.400881e+05  ...     1986.552492     12768.455652
std    2.876566e+09  3.671272e+05  ...      685.391304     27304.179631
min    1.000102e+06  7.500000e+04  ...      399.000000      651.000000
25%    2.123049e+09  3.219500e+05  ...     1490.000000     5100.000000
50%    3.904930e+09  4.500000e+05  ...     1840.000000     7620.000000
75%    7.308900e+09  6.450000e+05  ...     2360.000000    10083.000000
max    9.900000e+09  7.700000e+06  ...     6210.000000    871200.000000

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-----2-----
      id      price  ...  sqft_living15  sqft_lot15
count  2.161300e+04  2.161300e+04  ...   21613.000000   21613.000000
mean   4.580302e+09  5.400881e+05  ...    1986.552492   12768.455652
std    2.876566e+09  3.671272e+05  ...     685.391304   27304.179631
min    1.000102e+06  7.500000e+04  ...     399.000000     651.000000
25%    2.123049e+09  3.219500e+05  ...    1490.000000     5100.000000
50%    3.904930e+09  4.500000e+05  ...    1840.000000     7620.000000
75%    7.308900e+09  6.450000e+05  ...    2360.000000    10083.000000
max    9.900000e+09  7.700000e+06  ...    6210.000000   871200.000000

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[8 rows x 20 columns]

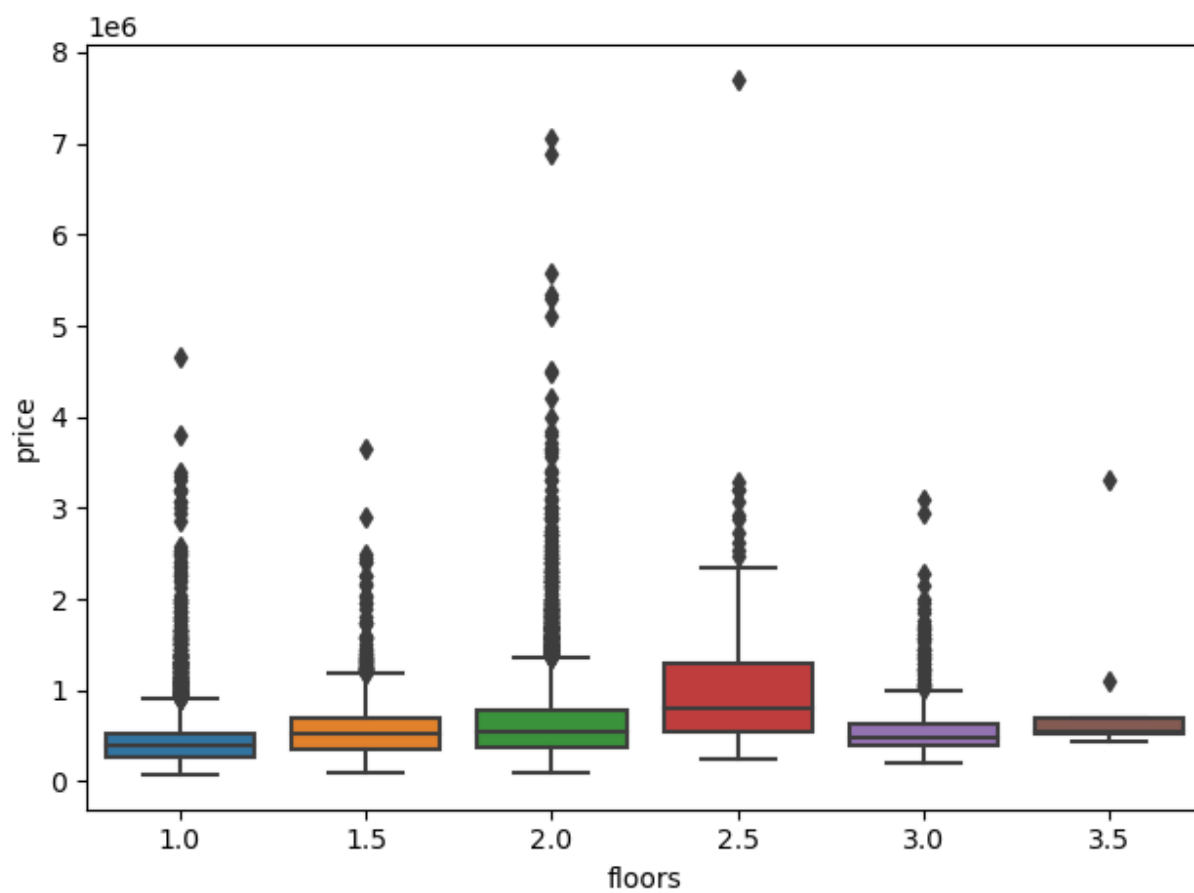
```

-----3-----
      floors
1.0    10680
2.0     8241
1.5     1910
3.0      613
2.5     161
3.5       8

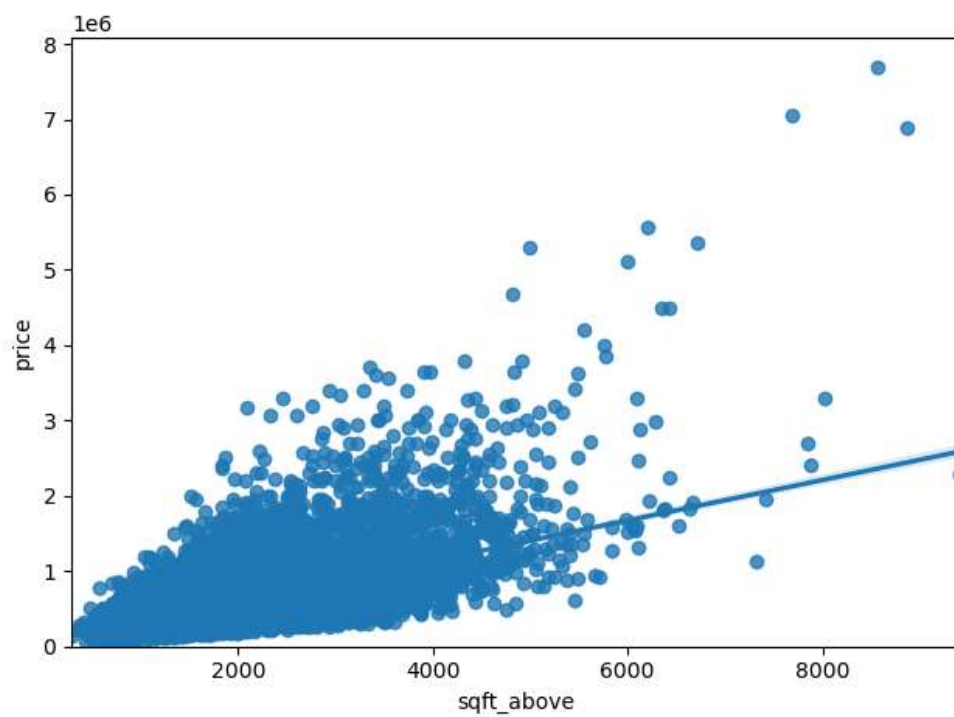
-----4-----
-----5-----
-----6-----
0.4928532179037931
-----7-----
0.6577151058279325
-----8-----
0.7513110900226091
-----9-----
number of test samples : 3242
number of training samples: 18371
-----10-----
0.7004432064921222

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-----4-----



-----5-----



```
-----6-----  
0.4928532179037931  
-----7-----  
0.6577151058279325  
-----8-----  
0.7513110900226091  
-----9-----  
number of test samples : 3242  
number of training samples: 18371  
-----10-----  
0.7004432064921222
```