Import necessary libraries

```
In [1]: import pandas as pd
   import sklearn
   import numpy as np
   import seaborn as sns
   import matplotlib
   import matplotlib.pyplot as plt

%matplotlib inline
```

```
Read Files
In [2]: | filename = 'C:/Users/hoags/Desktop/Dataset/HouseData/House-Dataset.xlsx'
        df = pd.read_excel(filename)
        df.shape
Out[2]: (21060, 16)
In [3]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 21060 entries, 0 to 21059
        Data columns (total 16 columns):
             Column
                          Non-Null Count
                                          Dtype
             -----
                          _____
                                          ____
         0
             id
                          21060 non-null
                                          int64
         1
             date
                          21060 non-null
                                          object
         2
             price
                          21060 non-null
                                          int64
         3
             bedrooms
                          21060 non-null
                                          int64
         4
                          21060 non-null
                                          float64
             bathrooms
                                         int64
         5
             sqft living 21060 non-null
         6
             sqft_lot
                          21060 non-null int64
         7
             floors
                          21060 non-null
                                          float64
         8
             waterfront 21060 non-null
                                          object
         9
                          21060 non-null
             view
                                          object
         10 condition
                                          object
                         21060 non-null
                          21060 non-null
                                          int64
         11 grade
         12 yr_built
                          21060 non-null
                                          int64
         13 zipcode
                          21060 non-null int64
         14 lat
                          21060 non-null float64
         15 long
                          21060 non-null float64
        dtypes: float64(4), int64(8), object(4)
        memory usage: 2.6+ MB
In [4]: # Convert date column to datetime
        df['date'] = pd.to_datetime(df['date'])
```

```
In [5]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 21060 entries, 0 to 21059
        Data columns (total 16 columns):
             Column
                          Non-Null Count Dtype
         0
             id
                          21060 non-null
                                          int64
         1
             date
                          21060 non-null datetime64[ns]
         2
             price
                          21060 non-null int64
         3
             bedrooms
                          21060 non-null int64
         4
             bathrooms
                         21060 non-null float64
         5
             sqft living 21060 non-null int64
             sqft_lot
         6
                          21060 non-null int64
         7
             floors
                          21060 non-null float64
             waterfront
         8
                         21060 non-null object
         9
             view
                         21060 non-null
                                          object
         10 condition
                         21060 non-null
                                          object
         11 grade
                         21060 non-null int64
         12 yr built
                          21060 non-null int64
         13 zipcode
                         21060 non-null int64
         14 lat
                          21060 non-null float64
         15 long
                          21060 non-null float64
        dtypes: datetime64[ns](1), float64(4), int64(8), object(3)
        memory usage: 2.6+ MB
In [6]:
        # Check the unique value count for each column
        for column in df.columns:
            unique values count = df[column].nunique()
            print('Number of unique value in', column, ':', unique_values_count)
        Number of unique value in id : 20887
        Number of unique value in date : 372
        Number of unique value in price : 3748
        Number of unique value in bedrooms : 8
        Number of unique value in bathrooms : 26
        Number of unique value in sqft living: 942
        Number of unique value in sqft lot: 9518
        Number of unique value in floors : 6
        Number of unique value in waterfront : 2
        Number of unique value in view : 5
        Number of unique value in condition : 5
        Number of unique value in grade : 11
        Number of unique value in yr built : 116
        Number of unique value in zipcode : 70
        Number of unique value in lat: 5026
        Number of unique value in long : 752
```

In [7]: # Drop columns id, lat, long
df.drop(columns = ['id', 'lat', 'long'])

	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	С
0	2014- 10 - 22	1495000	4	3.25	3070	10375	2.0	No	No View	
1	2014- 11-26	1495000	4	3.75	3770	4000	2.5	No	No View	
2	2015- 01-29	1490000	4	2.50	2420	18480	1.0	No	No View	
3	2015- 03-31	1490000	3	3.50	4560	14608	2.0	No	Average	
4	2014- 09-11	1490000	5	3.50	3620	7821	2.0	No	Average	
21055	2014- 11 - 05	82000	3	1.00	860	10426	1.0	No	No View	
21056	2015- 03-24	81000	2	1.00	730	9975	1.0	No	No View	
21057	2014- 05-23	80000	1	0.75	430	5050	1.0	No	No View	
21058	2014 - 05 - 06	78000	2	1.00	780	16344	1.0	No	No View	
	1 2 3 4 21055 21056	0 2014- 10-22 1 2014- 11-26 2 2015- 01-29 3 2015- 03-31 4 2014- 09-11 21055 2014- 11-05 21056 2015- 03-24 21057 2014- 05-23	0 2014- 1495000 1 2014- 1495000 2 2015- 01-29 1490000 3 2015- 03-31 1490000 4 2014- 09-11 1490000 21055 2014- 11-05 82000 21056 2015- 03-24 81000 21057 2014- 05-23 80000 21058 2014- 78000	0 2014- 10-22 1495000 4 1 2014- 1495000 4 2 2015- 01-29 1490000 4 3 2015- 03-31 1490000 3 4 2014- 09-11 1490000 5 21055 2014- 11-05 82000 3 21056 2015- 03-24 81000 2 21057 2014- 05-23 80000 1 21058 2014- 78000 2	0 2014- 10-22 1495000 4 3.25 1 2014- 11-26 1495000 4 3.75 2 2015- 01-29 1490000 4 2.50 3 2015- 03-31 1490000 3 3.50 4 2014- 09-11 1490000 5 3.50 21055 2014- 11-05 82000 3 1.00 21056 2015- 03-24 81000 2 1.00 21057 2014- 05-23 80000 1 0.75 21058 2014- 78000 2 1.00	0 2014- 10-22 1495000 4 3.25 3070 1 2014- 1495000 4 3.75 3770 2 2015- 01-29 1490000 4 2.50 2420 3 2015- 03-31 1490000 3 3.50 4560 4 2014- 09-11 1490000 5 3.50 3620 21055 2014- 11-05 82000 3 1.00 860 21056 2015- 03-24 81000 2 1.00 730 21057 2014- 05-23 80000 1 0.75 430 21058 2014- 78000 2 1.00 780	0 2014- 10-22 1495000 4 3.25 3070 10375 1 2014- 11-26 1495000 4 3.75 3770 4000 2 2015- 01-29 1490000 4 2.50 2420 18480 3 2015- 03-31 1490000 3 3.50 4560 14608 4 2014- 09-11 1490000 5 3.50 3620 7821 21055 2014- 11-05 82000 3 1.00 860 10426 21056 2015- 03-24 81000 2 1.00 730 9975 21057 2014- 05-23 80000 1 0.75 430 5050	0 2014- 10-22 1495000 4 3.25 3070 10375 2.0 1 2014- 11-26 1495000 4 3.75 3770 4000 2.5 2 2015- 01-29 1490000 4 2.50 2420 18480 1.0 3 2015- 03-31 1490000 3 3.50 4560 14608 2.0 4 2014- 09-11 1490000 5 3.50 3620 7821 2.0 21055 2014- 11-05 82000 3 1.00 860 10426 1.0 21056 2015- 03-24 81000 2 1.00 730 9975 1.0 21057 2014- 05-23 80000 1 0.75 430 5050 1.0 21058 2014- 05-23 78000 2 1.00 780 16344 1.0	0 2014- 10-22 1495000 4 3.25 3070 10375 2.0 No 1 2014- 11-26 1495000 4 3.75 3770 4000 2.5 No 2 2015- 01-29 1490000 4 2.50 2420 18480 1.0 No 3 2015- 01-29 1490000 3 3.50 4560 14608 2.0 No 4 2014- 09-11 1490000 5 3.50 3620 7821 2.0 No 2014- 09-11 1490000 5 3.50 3620 7821 2.0 No 21055 2014- 11-05 82000 3 1.00 860 10426 1.0 No 21056 2015- 03-24 81000 2 1.00 730 9975 1.0 No 21057 2014- 05-23 80000 1 0.75 430 5050 1.0 No 21058 2014- 05-23 80000 1 0.75 430 5050 1.0 No	0 2014- 10-22 1495000 4 3.25 3070 10375 2.0 No No View 1 2014- 11-26 1495000 4 3.75 3770 4000 2.5 No No View 2 2015- 01-29 1490000 4 2.50 2420 18480 1.0 No No View 3 2015- 03-31 1490000 3 3.50 4560 14608 2.0 No Average 4 2014- 09-11 1490000 5 3.50 3620 7821 2.0 No Average <

0.00

670

43377

1.0

No No View

21060 rows × 13 columns

75000

1

2015-02**-**17

21059

```
In [8]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 21060 entries, 0 to 21059
Data columns (total 16 columns):
     Column
                  Non-Null Count
                                  Dtype
 0
     id
                  21060 non-null
                                  int64
     date
 1
                                  datetime64[ns]
                  21060 non-null
 2
     price
                  21060 non-null
                                  int64
 3
     bedrooms
                  21060 non-null
                                  int64
 4
     bathrooms
                  21060 non-null
                                  float64
 5
     sqft living 21060 non-null
                                  int64
 6
     sqft lot
                  21060 non-null
                                  int64
 7
     floors
                  21060 non-null
                                  float64
 8
     waterfront
                  21060 non-null
                                  object
 9
     view
                  21060 non-null
                                  object
 10
    condition
                  21060 non-null
                                  object
 11
    grade
                  21060 non-null
                                  int64
                  21060 non-null
 12 yr_built
                                  int64
 13 zipcode
                  21060 non-null
                                  int64
 14 lat
                  21060 non-null
                                  float64
 15 long
                  21060 non-null float64
dtypes: datetime64[ns](1), float64(4), int64(8), object(3)
memory usage: 2.6+ MB
```

Outlier Detection for Numerical Feature

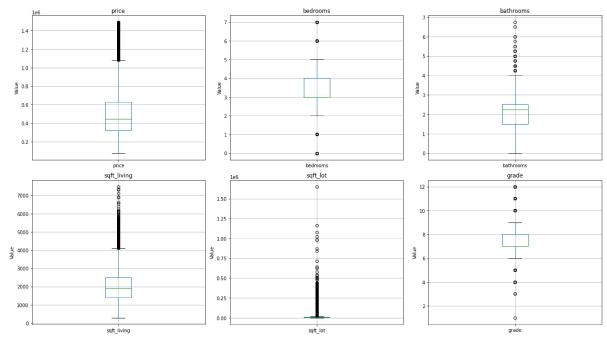
```
In [9]: def count_outliers(data, col):
            q1 = data[col].quantile(0.25, interpolation='nearest')
            q3 = data[col].quantile(0.75, interpolation='nearest')
            IQR = q3 - q1
            LLP = q1 - 1.5 * IQR
            ULP = q3 + 1.5 * IQR
            lower_outliers = data[data[col] < LLP][col].size</pre>
            upper outliers = data[data[col] > ULP][col].size
            if lower outliers == 0 and upper outliers ==0:
                print("No outliers in", col)
            else:
                print("There are outliers in", col)
                print("Count of lower outliers", lower outliers)
                print("Count of upper outliers", upper_outliers)
        numerical features = ['price', 'bedrooms', 'bathrooms', 'sqft living', 'sqft l
        for feature in numerical features:
            count outliers(df, feature)
```

There are outliers in price Count of lower outliers 0 Count of upper outliers 713 There are outliers in bedrooms Count of lower outliers 212 Count of upper outliers 280 There are outliers in bathrooms Count of lower outliers 0 Count of upper outliers 139 There are outliers in sqft living Count of lower outliers 0 Count of upper outliers 383 There are outliers in saft lot Count of lower outliers 0 Count of upper outliers 2312 No outliers in floors There are outliers in grade Count of lower outliers 275 Count of upper outliers 1230 No outliers in yr built No outliers in zipcode

```
In [10]: fig, axes = plt.subplots (nrows=2, ncols=3, figsize=(18,10))
    axes = axes.flatten()

outliers_numerical = ['price', 'bedrooms', 'bathrooms', 'sqft_living', 'sqft_left
    # Plot box plots for numerical features
    for i, col in enumerate(outliers_numerical):
        ax = axes[i]
        df.boxplot(column=[col], ax=ax)
        ax.set_title(col)
        ax.set_ylabel('Value')

plt.tight_layout()
    plt.show()
```

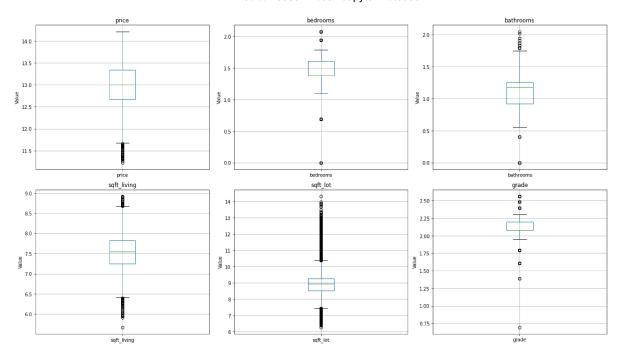


Transformation for numerical features

0 1 [44]	1]: price							
Out[11]:			bedrooms	bathrooms	sqft_living	sqft_lot	grade	
	0	1495000	4	3.25	3070	10375	10	
	1	1495000	4	3.75	3770	4000	9	
	2	1490000	4	2.50	2420	18480	9	
	3	1490000	3	3.50	4560	14608	12	
	4	1490000	5	3.50	3620	7821	10	

With log transformation

```
In [12]: for feature in outliers_numerical:
             data transformed[feature] = np.log1p(data transformed[feature])
         # Check outlier
         for feature in outliers numerical:
             count outliers(data transformed, feature)
         fig, axes = plt.subplots(nrows=2, ncols=3, figsize=(18,10))
         axes = axes.flatten()
         # Plot box plots for numerical features
         for i, col in enumerate(outliers numerical):
             ax = axes[i]
             data transformed.boxplot(column = [col], ax=ax)
             ax.set_title(col)
             ax.set_ylabel('Value')
         plt.tight layout()
         plt.show()
         C:\Users\hoags\AppData\Local\Temp\ipykernel 5560\4075392425.py:2: SettingWith
         CopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s
         table/user guide/indexing.html#returning-a-view-versus-a-copy (https://panda
         s.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-ver
         sus-a-copy)
           data_transformed[feature] = np.log1p(data_transformed[feature])
         There are outliers in price
         Count of lower outliers 61
         Count of upper outliers 0
         There are outliers in bedrooms
         Count of lower outliers 212
         Count of upper outliers 280
         There are outliers in bathrooms
         Count of lower outliers 14
         Count of upper outliers 23
         There are outliers in sqft living
         Count of lower outliers 57
         Count of upper outliers 19
         There are outliers in sqft lot
         Count of lower outliers 988
         Count of upper outliers 1566
         There are outliers in grade
         Count of lower outliers 275
         Count of upper outliers 1230
```



```
In [13]: # Rename columns
    for feature in outliers_numerical:
        data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
        data_transformed.drop(outliers_numerical, axis=1, inplace=True)
        data_transformed
```

C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:3: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://panda

s.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-ver sus-a-copy)

data_transformed[f'{feature}_log'] = np.log1p(data_transformed[feature])
C:\Users\hoags\AppData\Local\Temp\ipykernel_5560\2303374318.py:4: SettingWith
CopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

data_transformed.drop(outliers_numerical, axis=1, inplace=True)

Out[13]:

	price_log	bedrooms_log	bathrooms_log	sqft_living_log	sqft_lot_log	grade_log
0	2.722455	0.959135	0.894830	2.200526	2.327009	1.223156
1	2.722455	0.959135	0.939282	2.223011	2.229401	1.194706
2	2.722235	0.959135	0.812157	2.173835	2.381812	1.194706
3	2.722235	0.869742	0.917920	2.243397	2.359853	1.271150
4	2.722235	1.026672	0.917920	2.218606	2.299048	1.223156
21055	2.510776	0.869742	0.526589	2.048737	2.327488	1.080418
21056	2.509779	0.741276	0.526589	2.027413	2.323166	1.026672
21057	2.508769	0.526589	0.444440	1.955310	2.254166	0.959135
21058	2.506707	0.741276	0.526589	2.036087	2.370400	1.026672
21059	2.503504	0.526589	0.000000	2.016072	2.457682	0.869742

21060 rows × 6 columns

One hot Encoding for categorical feature

```
In [14]: categorical_features = ['waterfront', 'view', 'condition']
         categorical dummies = pd.get dummies(df[categorical features], drop first=True
         categorical_dummies.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 21060 entries, 0 to 21059
         Data columns (total 9 columns):
              Column
                                         Non-Null Count Dtype
         ---
              ----
                                         -----
          0
              waterfront Yes
                                         21060 non-null uint8
          1
              view Excellent
                                         21060 non-null uint8
          2
              view_Fair
                                         21060 non-null uint8
          3
              view Good
                                         21060 non-null uint8
          4
              view_No View
                                         21060 non-null uint8
          5
              condition_Fair- Badly worn 21060 non-null uint8
          6
              condition Good
                                         21060 non-null uint8
          7
              condition Poor- Worn out
                                         21060 non-null uint8
          8
              condition_Very Good
                                         21060 non-null uint8
         dtypes: uint8(9)
         memory usage: 185.2 KB
```

One hot Encoding for datetime feature

```
In [15]: # import datetime
         # import hashlib
         # def encode datetime(col):
               # Convert the datetime object to a string
               dt string = col.dt.strftime('%Y-%m-%d')
               # Hash the string
               hash value = [hashlib.sha256(value.encode()).hexdigest() for value in dt
               # Convert the hash value to a numerical representation
               numerical_representation = [int(hash_value, 16)/ 1e16 for hash_value in |
               return numerical representation
         # datetime encoded = encode datetime(df['date'])
         # datetime encoded df = pd.DataFrame(datetime encoded, columns=["date"])
         # datetime_encoded_df
         df['year'] = df['date'].dt.year
         df['month'] = df['date'].dt.month
         df['day'] = df['date'].dt.day
         datetime encoded = df[['year','month','day']]
         datetime_encoded
```

Out[15]:

	year	month	day
0	2014	10	22
1	2014	11	26
2	2015	1	29
3	2015	3	31
4	2014	9	11
21055	2014	11	5
21056	2015	3	24
21057	2014	5	23
21058	2014	5	6
21059	2015	2	17

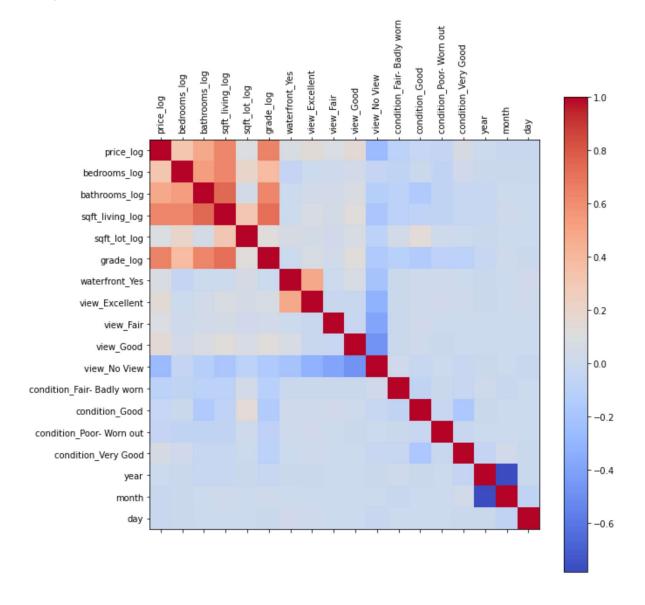
21060 rows × 3 columns

Final Dataset

Concat to datafinal

View Correlation

Out[18]: <matplotlib.colorbar.Colorbar at 0x124aca4aa60>



Training Model

```
In [19]: # Check null values
         datafinal.isnull().sum()
Out[19]: price_log
                                         0
                                         0
         bedrooms_log
         bathrooms log
                                         0
         sqft_living_log
                                         0
         sqft_lot_log
                                         0
         grade_log
                                         0
         waterfront Yes
                                         0
         view Excellent
                                         0
         view_Fair
                                         0
         view_Good
                                         0
         view_No View
                                         0
         condition_Fair- Badly worn
         condition Good
         condition Poor- Worn out
                                         0
         condition_Very Good
                                         0
         year
                                         0
         month
                                         0
         day
         dtype: int64
```

Split to Train and Test

```
In [20]: from sklearn.model_selection import train_test_split

# Feature selection
X = datafinal.drop(['price_log'], axis=1)
y = datafinal['price_log']

# Split train, test
X_train, X_test, y_train, y_test = train_test_split (X, y, test_size=0.2, random)
```

Linear Regression

```
In [21]: from sklearn.linear_model import LinearRegression
    from sklearn.metrics import mean_squared_error, r2_score

model_lr = LinearRegression()
    model_lr.fit(X_train, y_train)
    y_pred_lr = model_lr.predict(X_test)

mse_lr = mean_squared_error(y_test, y_pred_lr)
    r2_lr = r2_score(y_test, y_pred_lr)

print("Mean Squared Error:", mse_lr)
    print("R-squared:", r2_lr)
```

Mean Squared Error: 0.0005486938155256882

R-squared: 0.5189553959110034

Random Forest

```
In [22]: from sklearn.ensemble import RandomForestRegressor

model_rfc = RandomForestRegressor(random_state=42, n_estimators=200)
model_rfc.fit(X_train, y_train)
y_pred_rfc = model_rfc.predict(X_test)

mse_rfc = mean_squared_error(y_test, y_pred_rfc)
r2_rfc = r2_score(y_test, y_pred_rfc)

print("Mean Squared Error:", mse_rfc)
print("R-squared:", r2_rfc)
```

Mean Squared Error: 0.0005042231499933552 R-squared: 0.5579432122290711

In []: