dmv4

November 3, 2024

1 Data Wrangling Problem Statement: Data Wrangling on Real Estate Market The goal is to perform data wrangling to gain insights into the factors influencing housing prices and prepare the dataset for further analysis or modeling.

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

[2]:
```
df=pd.read_csv(r"C:\Users\dell\Desktop\DMV and ML\DMV Datasets\Real-Estate␣
↪dataset.csv")
```

[3]:
```
df.head()
```

[3]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished

[4]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 13 columns):
 #   Column            Non-Null Count  Dtype
---  ------            --------------  -----
```

```
 0   price             545 non-null    int64
 1   area              545 non-null    int64
 2   bedrooms          545 non-null    int64
 3   bathrooms         545 non-null    int64
 4   stories           545 non-null    int64
 5   mainroad          545 non-null    object
 6   guestroom         545 non-null    object
 7   basement          545 non-null    object
 8   hotwaterheating   545 non-null    object
 9   airconditioning   545 non-null    object
 10  parking           545 non-null    int64
 11  prefarea          545 non-null    object
 12  furnishingstatus  545 non-null    object
dtypes: int64(6), object(7)
memory usage: 55.5+ KB
```

[5]:
```
df.describe()
```

[5]:

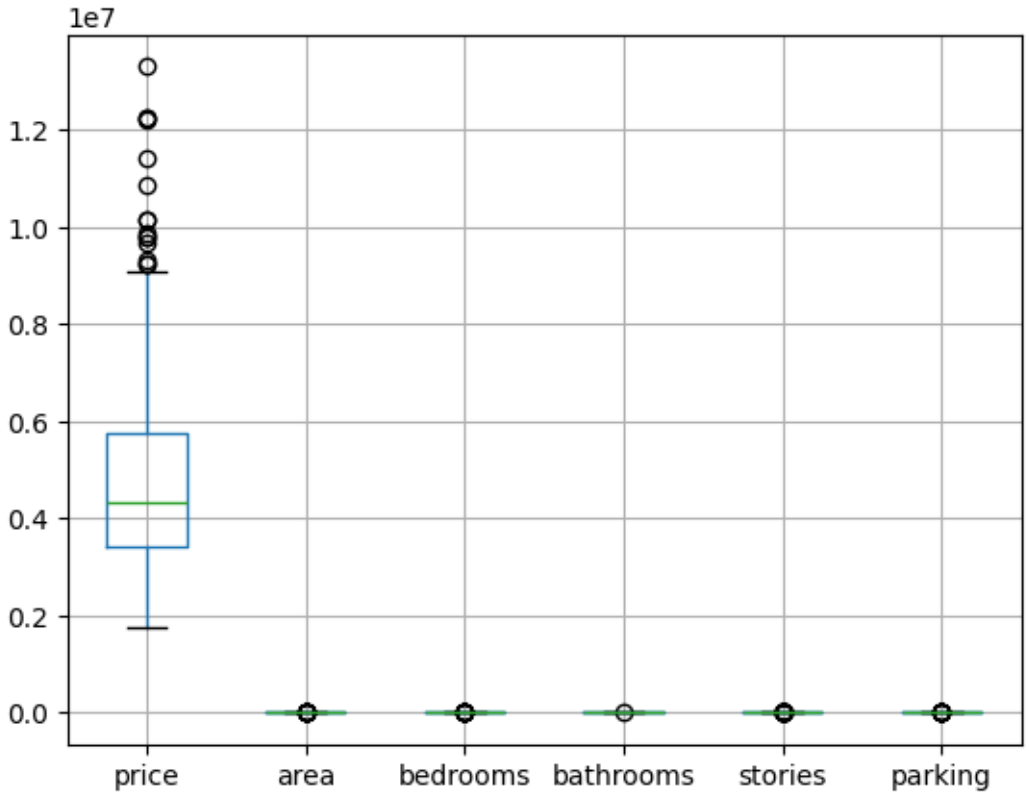
	price	area	bedrooms	bathrooms	stories	parking
count	5.450000e+02	545.000000	545.000000	545.000000	545.000000	545.000000
mean	4.766729e+06	5150.541284	2.965138	1.286239	1.805505	0.693578
std	1.870440e+06	2170.141023	0.738064	0.502470	0.867492	0.861586
min	1.750000e+06	1650.000000	1.000000	1.000000	1.000000	0.000000
25%	3.430000e+06	3600.000000	2.000000	1.000000	1.000000	0.000000
50%	4.340000e+06	4600.000000	3.000000	1.000000	2.000000	0.000000
75%	5.740000e+06	6360.000000	3.000000	2.000000	2.000000	1.000000
max	1.330000e+07	16200.000000	6.000000	4.000000	4.000000	3.000000

[6]:
```
df.isna().sum()
```

[6]:
```
price             0
area              0
bedrooms          0
bathrooms         0
stories           0
mainroad          0
guestroom         0
```

```
basement            0
hotwaterheating     0
airconditioning     0
parking             0
prefarea            0
furnishingstatus    0
dtype: int64
```

[7]:
```
df.boxplot()
plt.show()
```

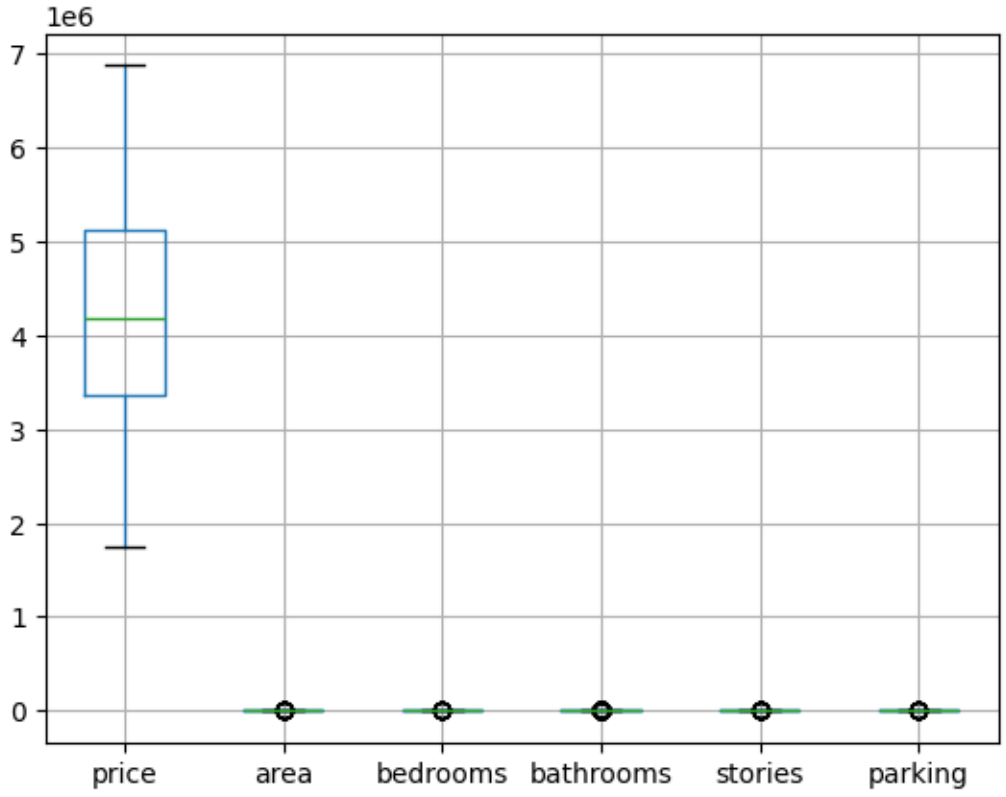


[8]:
```
Q1 = df["price"].quantile(0.25)
Q3 = df["price"].quantile(0.75)
iqr= Q3 - Q1
minm = Q1 - (1.5 * iqr)
maxm = Q3 + (1.5 * iqr)
df = df[(df["price"]>minm) & (df["price"]<maxm)]


# another method for same results
upper_limit = df['price'].quantile(0.9)
```

```
df = df[df['price'] < upper_limit]
```

[9]:
```
df.boxplot()
plt.show()
```



[10]:
```
df.dtypes
```

[10]:
```
price              int64
area               int64
bedrooms           int64
bathrooms          int64
stories            int64
mainroad          object
guestroom         object
basement          object
hotwaterheating   object
airconditioning   object
parking            int64
prefarea          object
furnishingstatus  object
dtype: object
```

[11]:
```
df.columns
```

[11]:
```
Index(['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad',
       'guestroom', 'basement', 'hotwaterheating', 'airconditioning',
       'parking', 'prefarea', 'furnishingstatus'],
      dtype='object')
```

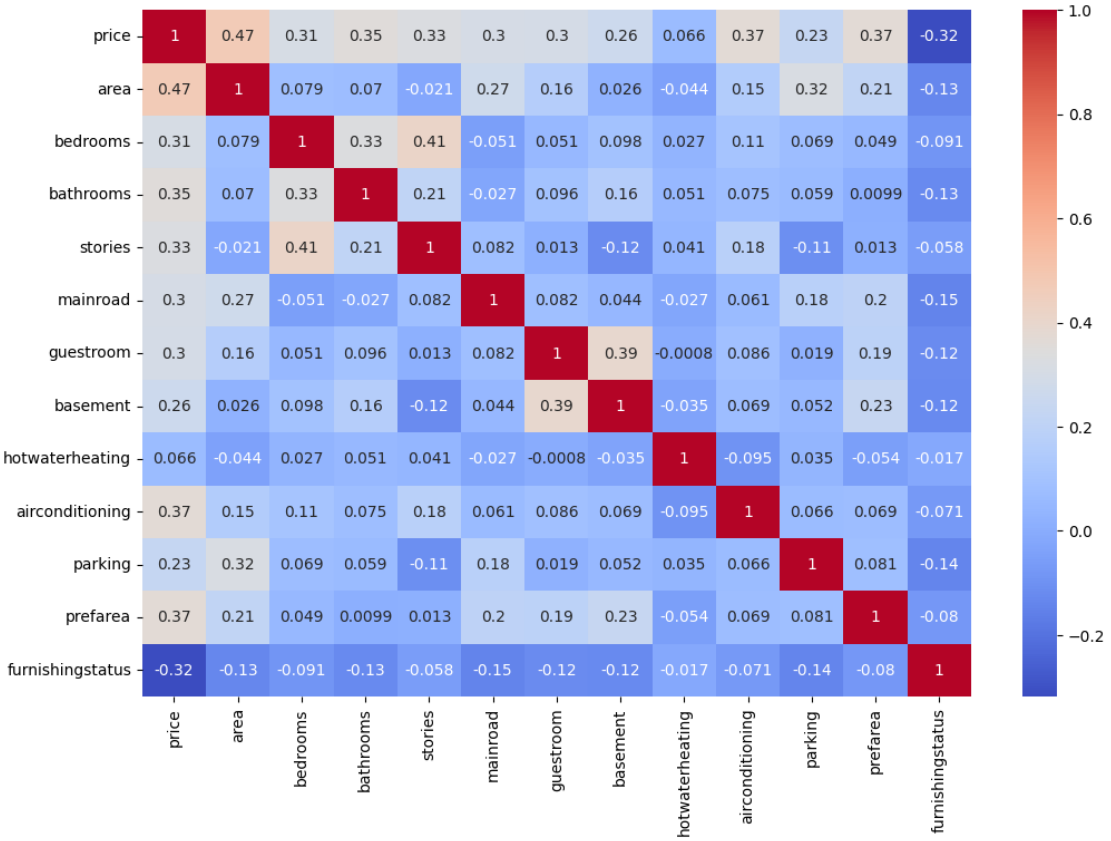
[12]:
```
le =LabelEncoder()
df['mainroad'] = le.fit_transform(df['mainroad'])
df['guestroom'] = le.fit_transform(df['guestroom'])
df['basement'] = le.fit_transform(df['basement'])
df['hotwaterheating'] = le.fit_transform(df['hotwaterheating'])
df['airconditioning'] = le.fit_transform(df['airconditioning'])
df['prefarea'] = le.fit_transform(df['prefarea'])
df['furnishingstatus'] = le.fit_transform(df['furnishingstatus'])
```

[13]:
```
df.head()
```

[13]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
68	6860000	6000	3	1	1	1	0	0	0	1	1	0	0
69	6790000	12090	4	2	2	1	0	0	0	0	2	1	0
70	6790000	4000	3	2	2	1	0	1	0	1	0	1	1
71	6755000	6000	4	2	4	1	0	0	0	1	0	0	2
72	6720000	5020	3	1	4	1	0	0	0	1	0	1	2

[14]:
```
plt.figure(figsize=(12, 8))
sns.heatmap(data = df.corr(), cmap = "coolwarm", annot=True)
plt.show()
```


price 1 0.47 0.31 0.35 0.33 0.3 0.3 0.26 0.066 0.37 0.23 0.37 -0.32
area 0.47 1 0.079 0.07 -0.021 0.27 0.16 0.026 -0.044 0.15 0.32 0.21 -0.13
bedrooms 0.31 0.079 1 0.33 0.41 -0.051 0.051 0.098 0.027 0.11 0.069 0.049 -0.091
bathrooms 0.35 0.07 0.33 1 0.21 -0.027 0.096 0.16 0.051 0.075 0.059 0.0099 -0.13
stories 0.33 -0.021 0.41 0.21 1 0.082 0.013 -0.12 0.041 0.18 -0.11 0.013 -0.058
mainroad 0.3 0.27 -0.051 -0.027 0.082 1 0.082 0.044 -0.027 0.061 0.18 0.2 -0.15
guestroom 0.3 0.16 0.051 0.096 0.013 0.082 1 0.39 -0.0008 0.086 0.019 0.19 -0.12
basement 0.26 0.026 0.098 0.16 -0.12 0.044 0.39 1 -0.035 0.069 0.052 0.23 -0.12
hotwaterheating 0.066 -0.044 0.027 0.051 0.041 -0.027 -0.0008 -0.035 1 -0.095 0.035 -0.054 -0.017
airconditioning 0.37 0.15 0.11 0.075 0.18 0.061 0.086 0.069 -0.095 1 0.066 0.069 -0.071
parking 0.23 0.32 0.069 0.059 -0.11 0.18 0.019 0.052 0.035 0.066 1 0.081 -0.14
prefarea 0.37 0.21 0.049 0.0099 0.013 0.2 0.19 0.23 -0.054 0.069 0.081 1 -0.08
furnishingstatus -0.32 -0.13 -0.091 -0.13 -0.058 -0.15 -0.12 -0.12 -0.017 -0.071 -0.14 -0.08 1
price area bedrooms bathrooms stories mainroad guestroom basement hotwaterheating airconditioning parking prefarea furnishingstatus
1.0 0.8 0.6 0.4 0.2 0.0 −0.2

[ ]: