import pandas as pd In [162... import matplotlib.pyplot as plt import seaborn as sns from sklearn.preprocessing import OneHotEncoder from sklearn.metrics import mean\_absolute\_error from sklearn.model\_selection import train\_test\_split from sklearn.neighbors import KNeighborsClassifier from sklearn.metrics import accuracy\_score from sklearn import svm from sklearn.svm import SVC from sklearn.metrics import mean\_absolute\_percentage\_error from sklearn.linear\_model import LinearRegression dataset = pd.read\_csv('data.csv') dataset In [139... Out[139]: price bedrooms bathrooms sqft\_living sqft\_lot floors waterfront view condition sqft\_above sqft\_basement yr\_built yr\_renovated date city statezip country street 2014-05-02 3.130000e+05 18810 Densmore Ave WA USA 0 3.0 1.50 1340 7912 1.5 1340 1955 3 Shoreline 0 98133 2014-05-02 2.384000e+06 WA 5.0 2.50 3650 9050 2.0 3370 280 1921 709 W Blaine St USA 5 Seattle 00:00:00 98119 2014-05-02 26206-26214 143rd WA 3.420000e+05 USA 2 Kent 3.0 2.00 1930 11947 1.0 1930 1966 0 00:00:00 98042 Ave SE 2014-05-02 4.200000e+05 WA 3 8030 USA 2.25 2000 3.0 1.0 1000 1000 1963 857 170th Pl NE Bellevue 0 98008 2014-05-02 WA 5.500000e+05 4 4.0 2.50 1940 10500 1.0 1140 800 1976 1992 9105 170th Ave NE Redmond USA 0 00:00:00 98052 2014-07-09 00:00:00 WA 3.081667e+05 6360 1954 1979 Seattle 4595 3.0 1.75 1510 1.0 1510 501 N 143rd St USA 0 98133 2014-07-09 WA 2009 5.343333e+05 USA 2.50 1460 7573 2.0 1460 1983 14855 SE 10th Pl 4596 3.0 3 Bellevue 0 0 00:00:00 98007 2014-07-09 4597 4.169042e+05 2.50 2.0 2009 USA 3.0 3 3010 3010 7014 759 Ilwaco Pl NE Renton 00:00:00 2014-07-10 2.034000e+05 4598 2.00 2090 6630 1070 1974 USA 1.0 1020 5148 S Creston St Seattle 00:00:00 98178 2014-07-10 00:00:00 WA 4599 2.206000e+05 2.50 1490 8102 2.0 1990 USA 1490 3.0 18717 SE 258th St Covington 98042 4600 rows × 18 columns dataset.head(10) In [140... Out[140]: statezip country price bedrooms bathrooms sqft\_living sqft\_lot floors waterfront view condition sqft\_above sqft\_basement yr\_built yr\_renovated date city street 2014-05-02 WA USA 313000.0 3.0 1.50 7912 1.5 1340 1955 18810 Densmore Ave N 0 1340 0 3 Shoreline 00:00:00 98133 2014-05-02 00:00:00 WA 2384000.0 5.0 2.50 3650 9050 2.0 280 1921 709 W Blaine St USA 3370 0 5 0 Seattle 4 2014-05-02 WA 26206-26214 143rd Ave 1930 USA 342000.0 3.0 2.00 1930 1.0 Kent 2 11947 1966 0 0 0 00:00:00 98042 2014-05-02 WA USA 2.25 8030 1.0 1000 1963 857 170th PI NE 420000.0 3.0 2000 1000 3 0 0 0 Bellevue 00:00:00 98008 2014-05-02 WA 4.0 2.50 1.0 1140 1976 9105 170th Ave NE USA 550000.0 1940 10500 800 1992 Redmond 0 4 00:00:00 98052 2014-05-02 WA USA 5 880 2.0 1.00 880 6380 1.0 1938 1994 522 NE 88th St 490000.0 0 3 Seattle 0 0 00:00:00 2014-05-02 WA 335000.0 USA 2.00 1.0 2616 174th Ave NE 2.0 1350 2560 1350 1976 6 0 0 3 0 0 Redmond 00:00:00 98052 2014-05-02 Maple WA 482000.0 2.50 35868 2710 USA 2.0 23762 SE 253rd Pl 4.0 2710 1989 0 0 0 3 00:00:00 Valley 98038 2014-05-02 46611-46625 SE 129th WA 2.50 88426 1.0 1985 0 USA 452500.0 3.0 2430 1570 860 0 0 North Bend 00:00:00 98045 2014-05-02 WA 640000.0 USA 2.00 1.5 1520 0 1945 2010 6811 55th Ave NE 4.0 1520 6200 0 0 9 3 Seattle 00:00:00 98115 dataset.tail(10) In [141... price bedrooms bathrooms sqft\_living sqft\_lot floors waterfront view condition sqft\_above sqft\_basement yr\_built yr\_renovated 2014-07-08 00:00:00 WA 4590 380680.555556 4.0 2.50 8331 2620 13602 SE 186th Pl USA 1991 2620 2.0 Renton 0 0 98058 2014-07-08 3529 SW Webster WA USA 4591 396166.666667 3.0 1.75 1880 5752 1.0 0 0 940 940 1945 Seattle 00:00:00 98126 2014-07-08 00:00:00 WA Federal 252980.000000 USA 4592 4.0 2.50 2530 8169 2.0 0 2530 1993 37654 18th PI S 0 3 0 98003 Way 5703 Charlotte WA 4593 3.0 2.50 2538 4600 2.0 0 2538 2013 1923 Auburn USA 0 3 98092 Ave SE 2014-07-09 00:00:00 210614.285714 WA 26306 127th Ave USA 4594 1610 1994 3.0 2.50 1610 7223 2.0 Kent 0 0 3 0 98030 2014-07-09 00:00:00 308166.666667 WA USA 4595 3.0 1.75 1510 6360 1.0 1954 0 1510 501 N 143rd St 0 4 Seattle 98133 2014-07-09 00:00:00 WA 4596 534333.333333 3.0 2.50 1460 7573 2.0 1460 1983 Bellevue USA 0 2009 14855 SE 10th Pl 0 3 98007 2014-07-09 00:00:00 416904.166667 WA 4597 USA 3.0 2.50 3010 7014 2.0 0 3010 2009 759 Ilwaco Pl NE 0 3 Renton 98059 2014-07-10 203400.000000 WA 0 0 USA 4598 4.0 2.00 2090 6630 1.0 1070 1020 1974 0 5148 S Creston St 3 Seattle 00:00:00 98178 2014-07-10 00:00:00 WA 4599 220600.000000 3.0 2.50 1490 8102 2.0 0 0 1490 1990 0 18717 SE 258th St Covington USA 98042 dataset.shape (4600, 18) Out[142]: obj = (dataset.dtypes == 'object') In [143... object\_cols = list(obj[obj].index) print("Categorical variables:",len(object\_cols)) Categorical variables: 5 int\_ = (dataset.dtypes == 'int') In [144... num\_cols = list(int\_[int\_].index) print("Integer variables:",len(num\_cols)) Integer variables: 0 fl = (dataset.dtypes == 'float') In [145... fl\_cols = list(fl[fl].index) print("Float variables:",len(fl\_cols)) Float variables: 4 plt.figure(figsize=(12, 6)) In [146... sns.heatmap(dataset.corr(), cmap = 'BrBG', fmt = '.2f', linewidths = 2,annot = True) <AxesSubplot:> Out[146]: 1.0 0.20 0.02 1.00 0.33 0.43 0.05 0.15 0.14 0.23 0.03 0.37 0.21 price --0.03 bedrooms - 0.20 1.00 0.55 0.59 0.07 0.18 -0.00 0.11 0.03 0.48 0.33 0.14 -0.06 - 0.8 bathrooms - 0.33 1.00 0.76 -0.12 0.55 0.11 0.49 0.08 0.21 0.69 0.30 0.46 -0.22 sqft living - 0.43 -0.12 0.76 1.00 0.21 0.34 0.12 0.88 0.59 0.31 -0.06 0.45 0.29 - 0.6 sqft\_lot - 0.05 0.07 0.11 0.21 1.00 0.00 0.02 0.07 0.00 0.22 0.03 0.05 -0.02 floors - 0.15 0.18 0.49 0.34 0.00 1.00 0.02 0.03 -0.28 0.52 -0.26 0.47 -0.23 - 0.4 waterfront - 0.14 -0.00 0.08 0.12 0.02 0.02 1.00 0.36 0.00 0.08 0.10 -0.02 0.01 1.00 - 0.2 view - 0.23 0.11 0.21 0.31 0.07 0.06 -0.06 0.02 0.03 0.36 0.17 0.32 -0.12 -0.28 condition - 0.03 0.03 -0.06 0.00 1.00 -0.18 -0.19 0.00 0.06 0.20 -0.40 - 0.0 0.88 1.00 sqft\_above - 0.37 0.69 0.22 0.52 0.08 0.17 -0.18 -0.04 0.41 -0.16 0.48 -0.16 -0.26 0.20 1.00 sqft\_basement - 0.21 0.33 0.30 0.45 0.03 0.10 0.32 -0.04 0.04 - -0.2 1.00 -0.32 yr\_built - 0.02 0.14 0.46 0.29 0.05 0.47 -0.02 -0.40 0.41 -0.16 -0.06 yr\_renovated --0.03 -0.22 -0.02 0.01 0.02 -0.19 0.04 -0.32 1.00 sqft\_basement condition bathrooms unique\_values = [] In [147... for col in object\_cols: unique\_values.append(dataset[col].unique().size) plt.figure(figsize=(10,6)) plt.title('No. Unique values of Categorical Features') plt.xticks(rotation=90) sns.barplot(x=object\_cols,y=unique\_values) <AxesSubplot:title={'center':'No. Unique values of Categorical Features'}> Out[147]: No. Unique values of Categorical Features 4000 3000 2000 1000 street city country statezip dataset.drop(['date'], In [148... axis=1, inplace=True) In [149... dataset['price'] = dataset['price'].fillna( dataset['price'].mean()) new\_dataset = dataset.dropna() In [150... new\_dataset.isnull().sum() In [151... price Out[151]: bedrooms bathrooms sqft\_living sqft\_lot floors waterfront view condition sqft\_above sqft\_basement yr\_built yr\_renovated street city statezip country dtype: int64 from sklearn.preprocessing import OneHotEncoder s = (new\_dataset.dtypes == 'object') object\_cols = list(s[s].index) print("Categorical variables:") print(object\_cols) print('No. of. categorical features: ', len(object\_cols)) Categorical variables: ['street', 'city', 'statezip', 'country'] No. of. categorical features: 4 OH\_encoder = OneHotEncoder(sparse=False) In [153... OH\_cols = pd.DataFrame(OH\_encoder.fit\_transform(new\_dataset[object\_cols])) OH\_cols.index = new\_dataset.index OH\_cols.columns = OH\_encoder.get\_feature\_names() df\_final = new\_dataset.drop(object\_cols, axis=1) df\_final = pd.concat([df\_final, OH\_cols], axis=1) C:\Users\KONDAPAKA RISHIKA\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning: Function get\_feature\_names is deprecated; get\_feature\_names is deprecated in 1.0 and will be removed in 1.2. Please use get\_feature\_names\_out instead. warnings.warn(msg, category=FutureWarning) X = df\_final.drop(['price'], axis=1) Y = df\_final['price'] In [155... X Out[155]: x2\_WA x2\_WA x2\_WA x2\_WA x2\_WA x2\_WA x2\_WA x2\_WA x2\_WA bedrooms bathrooms sqft\_living sqft\_lot floors waterfront view condition sqft\_above sqft\_basement ... x3\_USA 98168 98198 98288 98354 98166 98177 98178 98188 98199 0 ... 7912 1340 0.0 0.0 1.5 1.0 3.0 1.50 1340 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0 0 3 280 ... 5.0 3650 9050 2.0 3370 0.0 0.0 0.0 0.0 0.0 0.0 1 2.50 0 0.0 0.0 0.0 1.0 5 2 2.00 1930 11947 1.0 1930 0 ... 0.0 0.0 0.0 0.0 0.0 0.0 0.0 3.0 0 0.0 0 0.0 1.0 1000 ... 3 0.0 0.0 1.0 3.0 2.25 2000 8030 1.0 0 1000 0.0 0.0 0.0 0.0 0.0 0.0 0 4 0.0 800 ... 0.0 10500 1.0 0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 4 4.0 2.50 1940 0 1140 1.0 1.75 6360 1.0 0 0 1510 0 ... 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1510 1.0 4595 3.0 4596 1460 0.0 2.50 1460 7573 2.0 0.0 0.0 4597 3.0 2.50 7014 2.0 3010 0 ... 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 3010 1.0 0 4598 4.0 2090 6630 1.0 1020 ... 0.0 0.0 1.0 0.0 0.0 0.0 1.0 0 0 1070 2.00 3 0.0 0.0 0.0 0 ... 0.0 8102 2.0 0 1490 0.0 0.0 0.0 0.0 0.0 0.0 0.0 4599 3.0 2.50 1490 0 0.0 1.0 4  $4600 \text{ rows} \times 4659 \text{ columns}$ In [156... Y 3.130000e+05 Out[156]: 2.384000e+06 3.420000e+05 4.200000e+05 5.500000e+05 4595 3.081667e+05 4596 5.343333e+05 4597 4.169042e+05 4598 2.034000e+05 2.206000e+05 4599 Name: price, Length: 4600, dtype: float64 In [157... X\_train, X\_test, y\_train, y\_test = train\_test\_split(X1, y1, test\_size=0.5, random\_state=42) Features1 = ['bedrooms', 'bathrooms', 'sqft\_living', 'sqft\_lot', 'floors'] In [158... target = 'price' X1 = dataset[Features1] y1 = dataset[target] In [159... X1 bedrooms bathrooms sqft\_living sqft\_lot floors Out[159]: 1.50 7912 1.5 0 3.0 1340 3650 9050 2.0 5.0 2.50 2 3.0 1.0 2.00 1930 11947 3 2000 8030 3.0 2.25 1.0 2.50 10500 4595 3.0 1.75 1510 6360 1.0 4596 1460 7573 2.0 3.0 2.50 4597 2.50 3010 7014 3.0 2.0 2090 6630 1.0 4598 4.0 2.00 3.0 2.50 1490 8102 2.0 4599 4600 rows  $\times$  5 columns In [160... y1 3.130000e+05 Out[160]: 2.384000e+06 3.420000e+05 4.200000e+05 5.500000e+05 4595 3.081667e+05 4596 5.343333e+05 4.169042e+05 4597 4598 2.034000e+05 2.206000e+05 4599 Name: price, Length: 4600, dtype: float64 model = LinearRegression() In [163... In [164... model.fit(X\_train, y\_train) LinearRegression() Out[164]: y\_pred = model.predict(X\_test) y\_pred In [166... array([363715.81566702, 399446.44586549, 834230.23979668, ..., Out[166]: 748123.84167884, 569569.52453247, 658083.4182335 ]) score = model.score(X\_test, y\_test) print("Model R^2 Score:", score) Model R^2 Score: 0.10933671026238057 new\_house = pd.DataFrame({'bedrooms': [2], 'bathrooms': [2.5], 'sqft\_living': [600], 'sqft\_lot': [600], 'floors': [2]}) In [169... predicted\_price = model.predict(new\_house) print("Predicted Price:", predicted\_price[0]) Predicted Price: 146977.0030232838 In [170... plt.figure(figsize=(18, 36)) plt.title('Categorical Features: Distribution') plt.xticks(rotation=90) index = 1for col in object\_cols: y = dataset[col].value\_counts() plt.subplot(11, 4, index) plt.xticks(rotation=90) sns.barplot(x=list(y.index), y=y) index += 1 150 1500 4000 125 1250 3 · 3000 statezip 1000 75 750 2000 -50 500 -1000 25 250 -COMMITTED ATTREET ATTR