

In [1]:

```

1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 from IPython import get_ipython
6 import warnings
7 warnings.filterwarnings("ignore")

```

In [2]:

```
1 diamond_data = pd.read_csv('diamonds.csv')
```

In [3]:

```
1 diamond_data.head()
```

Out[3]:

	Unnamed: 0	carat	cut	color	clarity	depth	table	price	x	y	z
0	1	0.23	Ideal	E	SI2	61.5	55.0	326	3.95	3.98	2.43
1	2	0.21	Premium	E	SI1	59.8	61.0	326	3.89	3.84	2.31
2	3	0.23	Good	E	VS1	56.9	65.0	327	4.05	4.07	2.31
3	4	0.29	Premium	I	VS2	62.4	58.0	334	4.20	4.23	2.63
4	5	0.31	Good	J	SI2	63.3	58.0	335	4.34	4.35	2.75

In [4]:

```
1 diamond_data.tail()
```

Out[4]:

	Unnamed: 0	carat	cut	color	clarity	depth	table	price	x	y	z
53935	53936	0.72	Ideal	D	SI1	60.8	57.0	2757	5.75	5.76	3.50
53936	53937	0.72	Good	D	SI1	63.1	55.0	2757	5.69	5.75	3.61
53937	53938	0.70	Very Good	D	SI1	62.8	60.0	2757	5.66	5.68	3.56
53938	53939	0.86	Premium	H	SI2	61.0	58.0	2757	6.15	6.12	3.74
53939	53940	0.75	Ideal	D	SI2	62.2	55.0	2757	5.83	5.87	3.64

In [8]:

```
1 diamond_data = diamond_data.drop('Unnamed: 0', axis = 1)
```

In [9]:



```
1 diamond_data.shape
```

Out[9]:

```
(53940, 10)
```

In [10]:



```
1 diamond_data.columns
```

Out[10]:

```
Index(['carat', 'cut', 'color', 'clarity', 'depth', 'table', 'price', 'x',  
'y',  
      'z'],  
      dtype='object')
```

In [11]:



```
1 diamond_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 53940 entries, 0 to 53939  
Data columns (total 10 columns):  
#   Column      Non-Null Count  Dtype  
---  -  
0   carat      53940 non-null  float64  
1   cut        53940 non-null  object  
2   color      53940 non-null  object  
3   clarity    53940 non-null  object  
4   depth      53940 non-null  float64  
5   table      53940 non-null  float64  
6   price      53940 non-null  int64  
7   x          53940 non-null  float64  
8   y          53940 non-null  float64  
9   z          53940 non-null  float64  
dtypes: float64(6), int64(1), object(3)  
memory usage: 4.1+ MB
```

In [12]:



```
1 diamond_data.describe()
```

Out[12]:

	carat	depth	table	price	x	y
count	53940.000000	53940.000000	53940.000000	53940.000000	53940.000000	53940.000000
mean	0.797940	61.749405	57.457184	3932.799722	5.731157	5.734526
std	0.474011	1.432621	2.234491	3989.439738	1.121761	1.142135
min	0.200000	43.000000	43.000000	326.000000	0.000000	0.000000
25%	0.400000	61.000000	56.000000	950.000000	4.710000	4.720000
50%	0.700000	61.800000	57.000000	2401.000000	5.700000	5.710000
75%	1.040000	62.500000	59.000000	5324.250000	6.540000	6.540000
max	5.010000	79.000000	95.000000	18823.000000	10.740000	58.900000

In [13]:



```
1 diamond_data.duplicated().sum()
```

Out[13]:

146

In [16]:



```
1 diamond_data = diamond_data.drop_duplicates()
```

In [17]:



```
1 diamond_data.shape
```

Out[17]:

(53794, 10)

In [18]:



```
1 diamond_data.isnull().sum()
```

Out[18]:

```
carat      0
cut         0
color       0
clarity     0
depth       0
table       0
price       0
x           0
y           0
z           0
dtype: int64
```

In [20]:



```
1 diamond_data.nunique()
```

Out[20]:

```
carat      273
cut         5
color       7
clarity     8
depth      184
table      127
price     11602
x          554
y          552
z          375
dtype: int64
```

In [21]:



```
1 diamond_data['cut'].unique()
```

Out[21]:

```
array(['Ideal', 'Premium', 'Good', 'Very Good', 'Fair'], dtype=object)
```

In [22]:



```
1 diamond_data['cut'].value_counts()
```

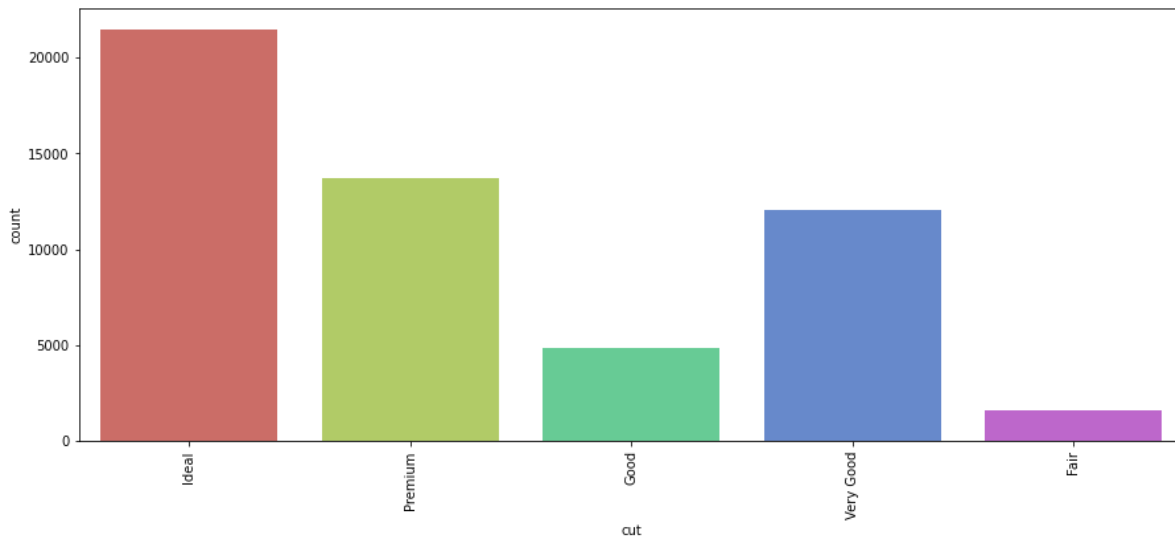
Out[22]:

```
Ideal      21488
Premium    13748
Very Good  12069
Good        4891
Fair        1598
Name: cut, dtype: int64
```

In [23]:



```
1 plt.figure(figsize=(15,6))
2 sns.countplot(diamond_data['cut'], data = diamond_data,
3               palette='hls')
4 plt.xticks(rotation = 90)
5 plt.show()
```



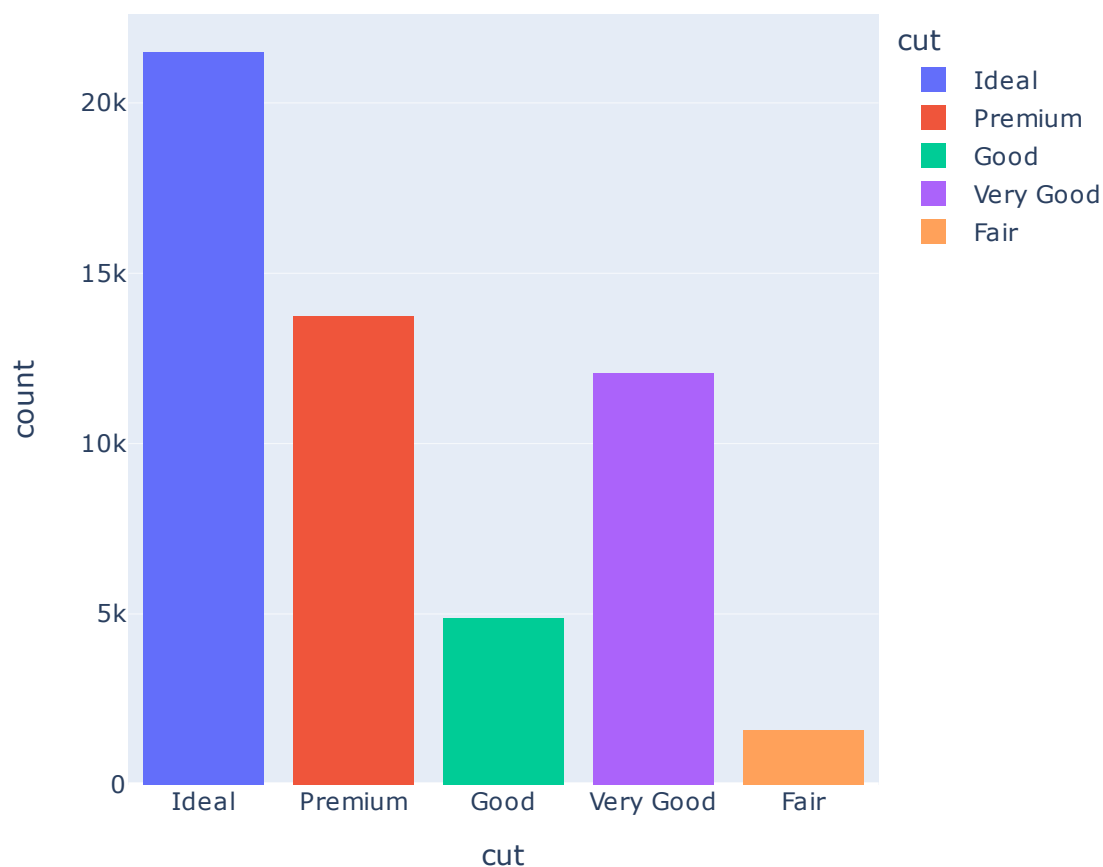
In [24]:



```
1 import plotly.express as px
```

In [25]:

```
1 fig1 = px.histogram(diamond_data, x = 'cut', color = 'cut')
2 fig1.show()
```



In [26]:

```
1 diamond_data['color'].unique()
```

Out[26]:

```
array(['E', 'I', 'J', 'H', 'F', 'G', 'D'], dtype=object)
```

In [27]:

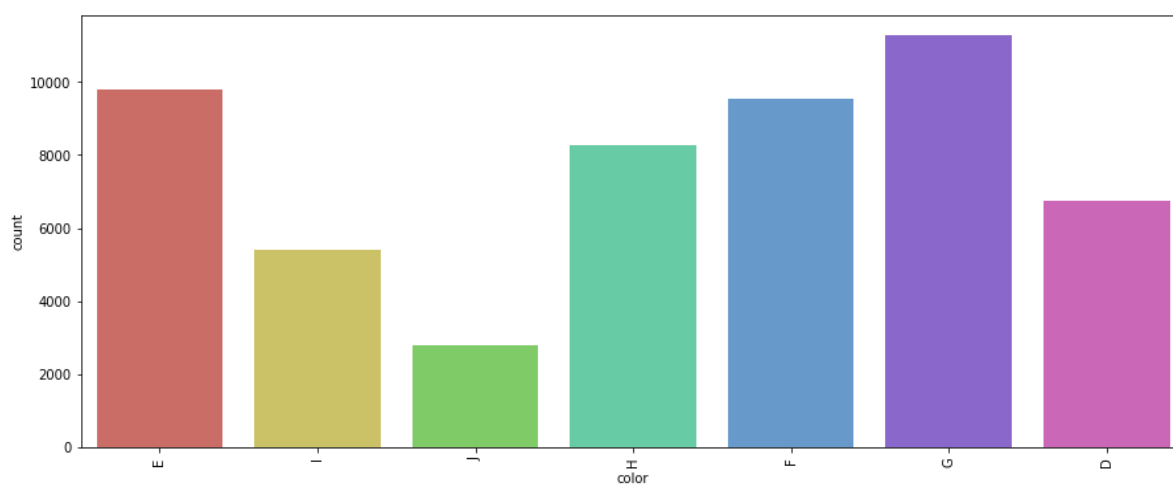
```
1 diamond_data['color'].value_counts()
```

Out[27]:

```
G      11262
E       9776
F       9520
H       8272
D       6755
I       5407
J       2802
Name: color, dtype: int64
```

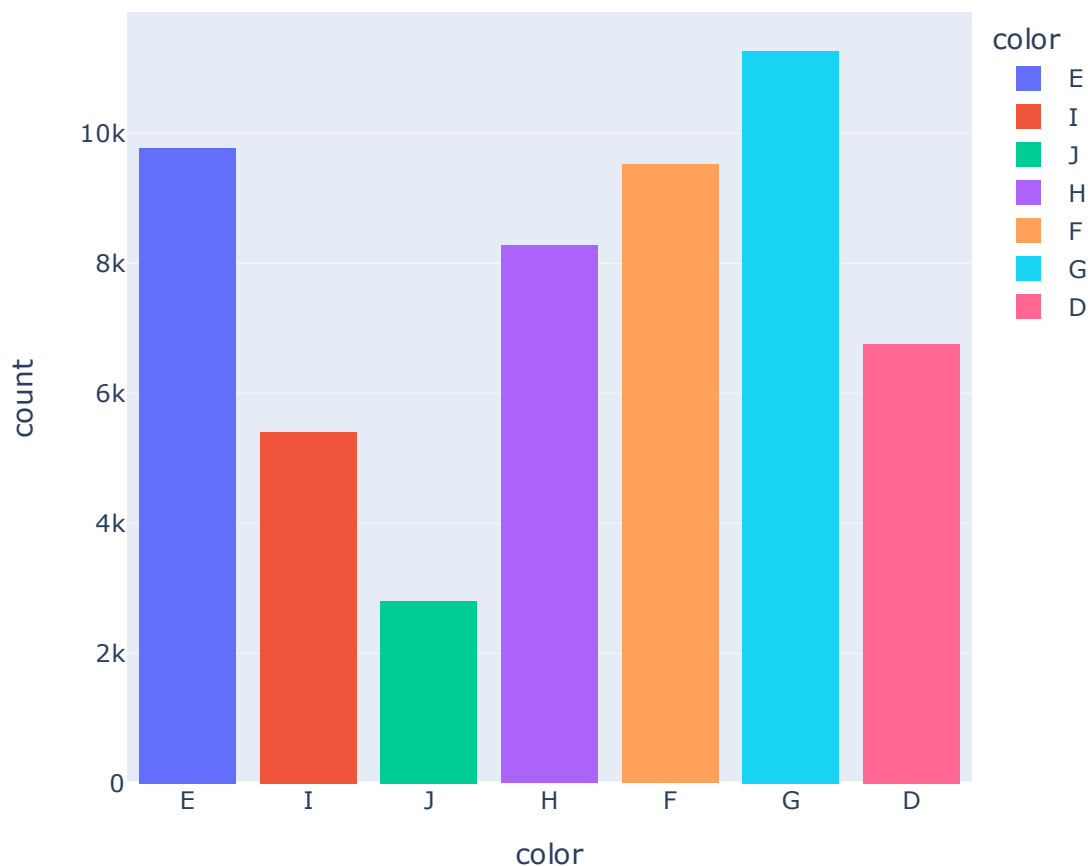
In [28]:

```
1 plt.figure(figsize=(15,6))
2 sns.countplot(diamond_data['color'], data = diamond_data,
3               palette='hls')
4 plt.xticks(rotation = 90)
5 plt.show()
```



In [29]:

```
1 fig2 = px.histogram(diamond_data, x = 'color', color = 'color')  
2 fig2.show()
```



In [30]:

```
1 diamond_data['clarity'].unique()
```

Out[30]:

```
array(['SI2', 'SI1', 'VS1', 'VS2', 'VVS2', 'VVS1', 'I1', 'IF'],  
      dtype=object)
```


In [31]:

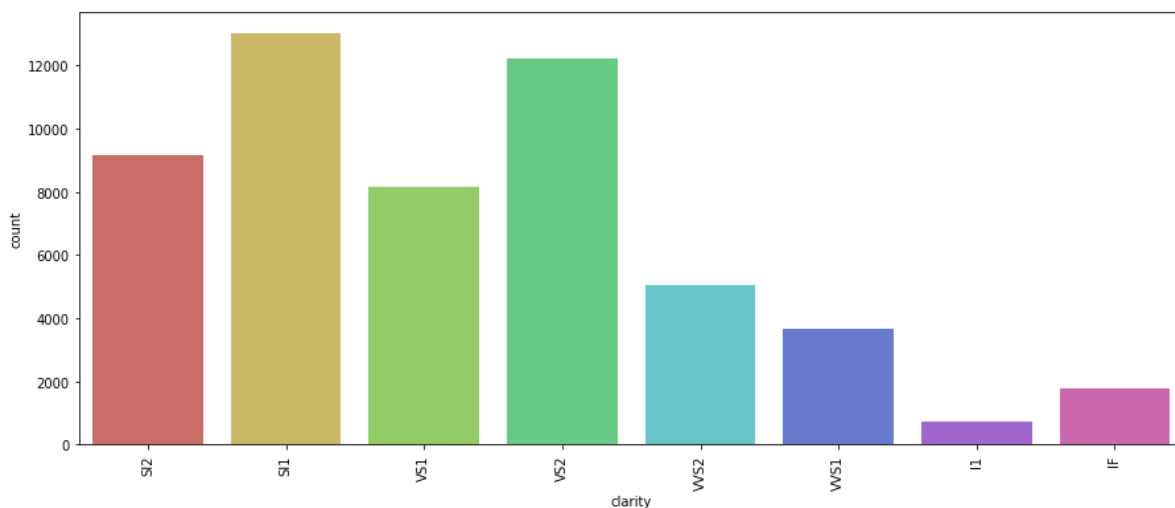
```
1 diamond_data['clarity'].value_counts()
```

Out[31]:

```
SI1    13032
VS2    12229
SI2     9150
VS1     8156
VVS2    5056
VVS1    3647
IF      1784
I1       740
Name: clarity, dtype: int64
```

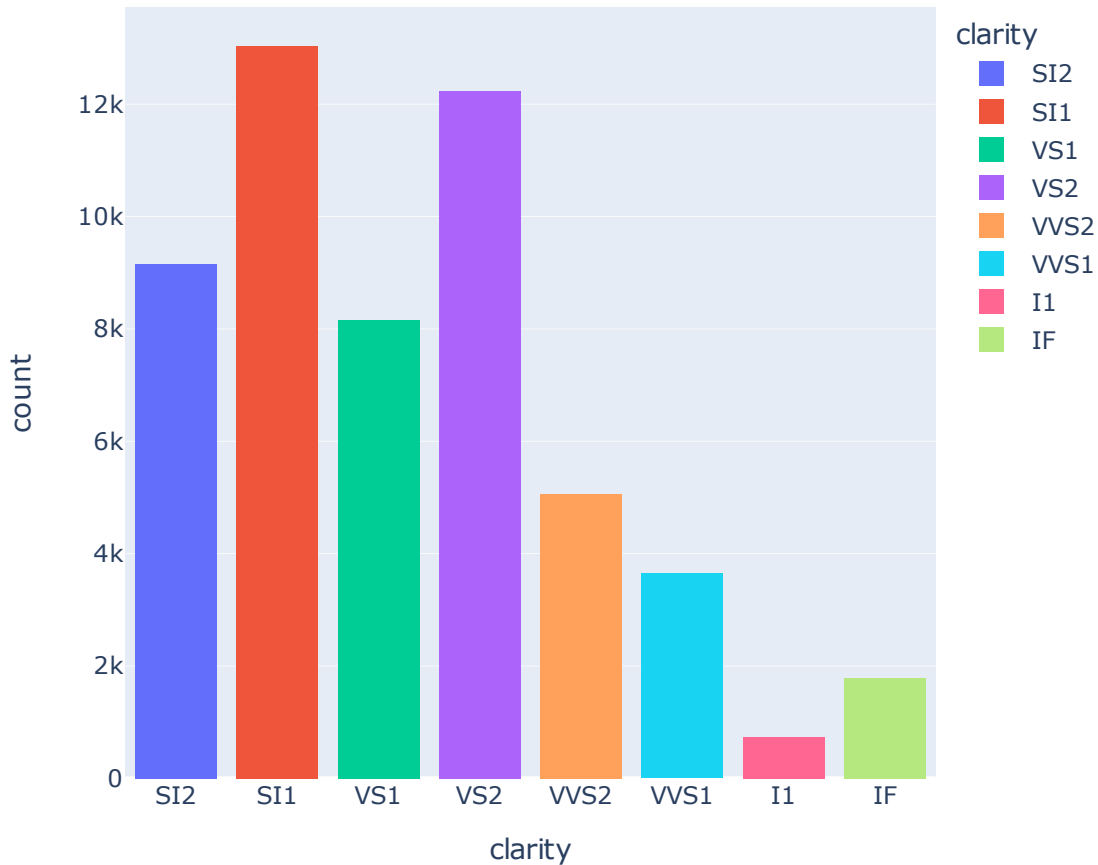
In [32]:

```
1 plt.figure(figsize=(15,6))
2 sns.countplot(diamond_data['clarity'], data = diamond_data,
3               palette='hls')
4 plt.xticks(rotation = 90)
5 plt.show()
```



In [34]:

```
1 fig3 = px.histogram(diamond_data, x = 'clarity', color = 'clarity')
2 fig3.show()
```



In [35]:

```
1 diamond_data = diamond_data.drop(diamond_data[diamond_data["x"]==0].index)
2 diamond_data = diamond_data.drop(diamond_data[diamond_data["y"]==0].index)
3 diamond_data = diamond_data.drop(diamond_data[diamond_data["z"]==0].index)
```

In [36]:

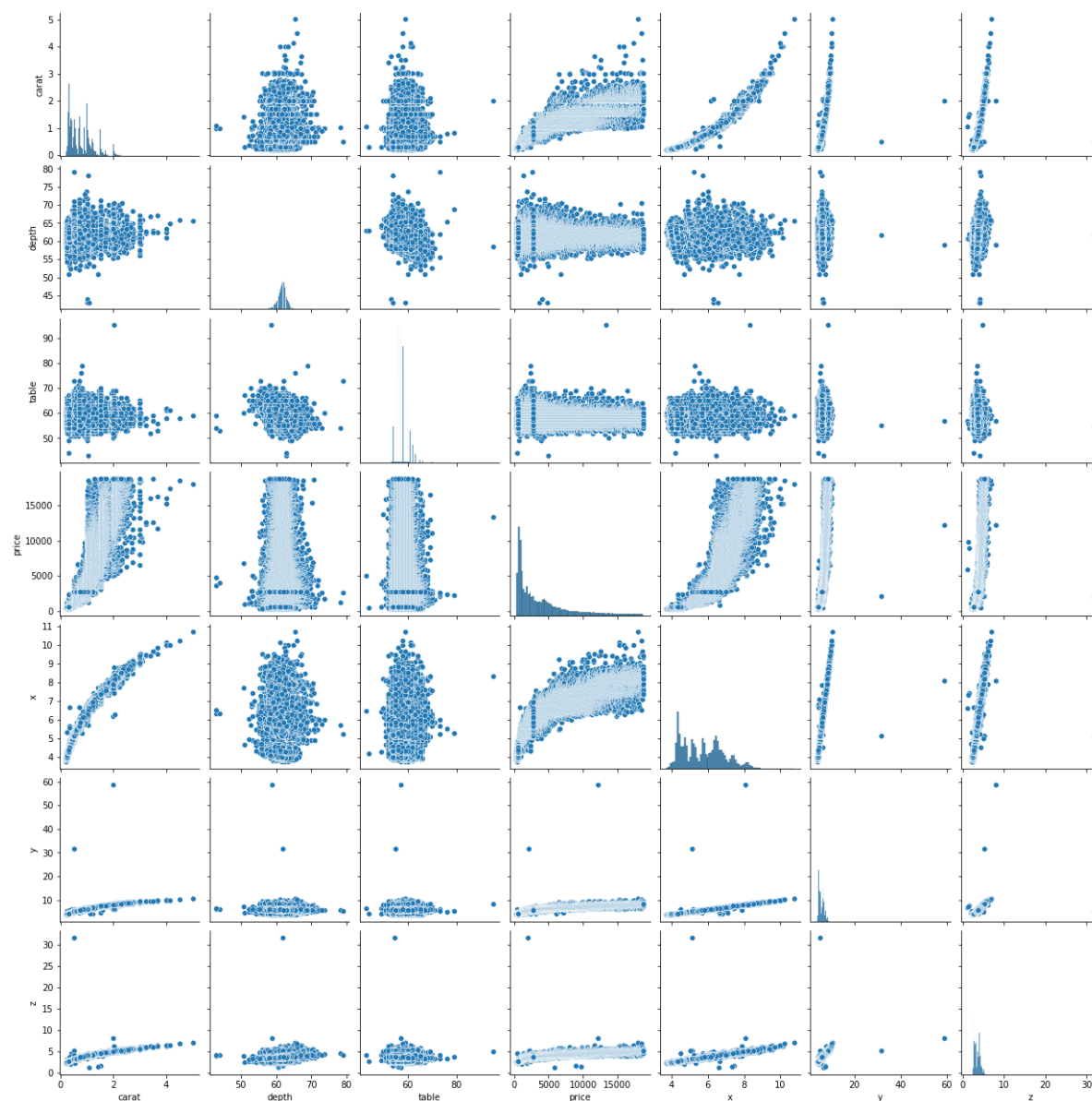
```
1 diamond_data.shape
```

Out[36]:

```
(53775, 10)
```

In [37]:

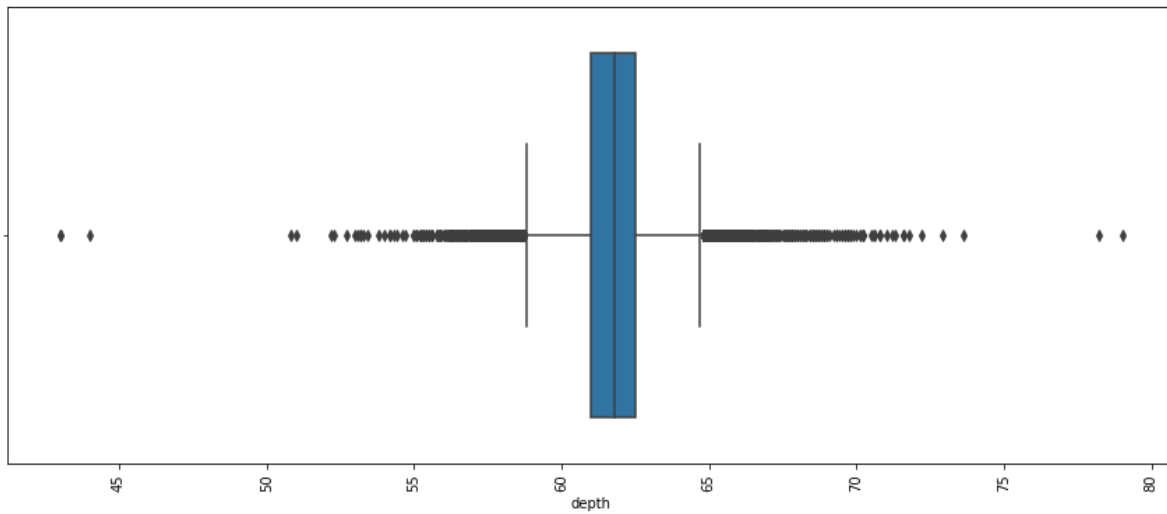
```
1 sns.pairplot(diamond_data)
2 plt.show()
```



In [38]:



```
1 plt.figure(figsize=(15,6))
2 sns.boxplot(diamond_data['depth'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [39]:

```

1 diamond_depth = diamond_data['depth']
2 Q3 = diamond_depth.quantile(0.75)
3 Q1 = diamond_depth.quantile(0.45)
4 IQR = Q3-Q1
5 lower_limit = Q1 -(1.5*IQR)
6 upper_limit = Q3 +(1.5*IQR)
7 depth_outliers = diamond_depth[(diamond_depth <lower_limit) | (diamond_depth >upper_limit)]
8 depth_outliers

```

Out[39]:

```

1      59.8
2      56.9
8      65.1
9      59.4
10     64.0
...
53918  59.3
53927  58.1
53930  60.5
53931  59.8
53932  60.5
Name: depth, Length: 11924, dtype: float64

```

In [40]:

```

diamond_data = diamond_data[(diamond_data["depth"]<75)&(diamond_data["depth"]>45)]

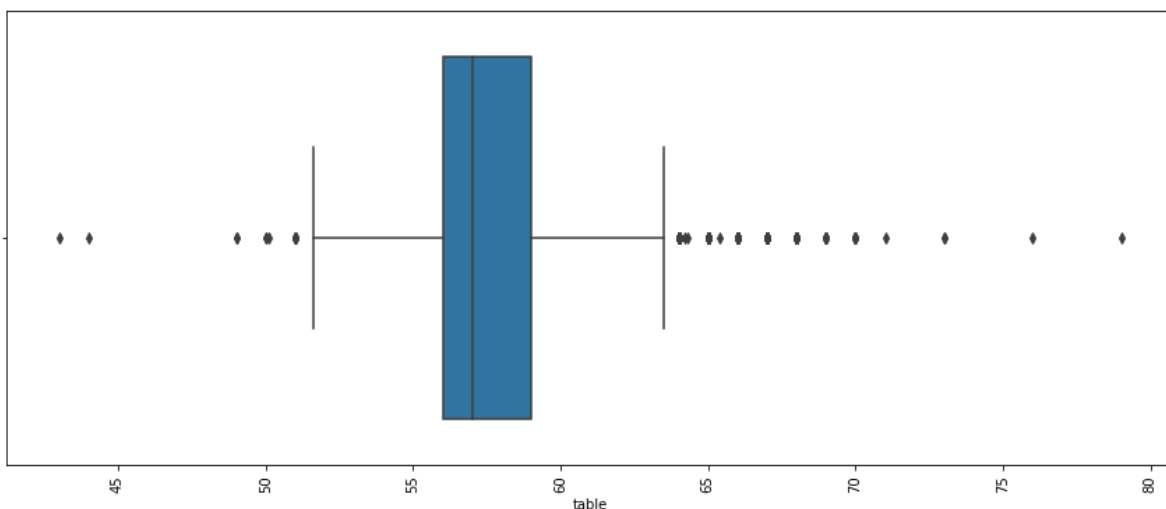
```

In [49]:

```

1 plt.figure(figsize=(15,6))
2 sns.boxplot(diamond_data['table'])
3 plt.xticks(rotation = 90)
4 plt.show()

```



In [41]:

```
1 diamond_table = diamond_data['table']
2 Q3 = diamond_table.quantile(0.80)
3 Q1 = diamond_table.quantile(0.40)
4 IQR = Q3-Q1
5 lower_limit = Q1 -(1.5*IQR)
6 upper_limit = Q3 +(1.5*IQR)
7 table_outliers = diamond_table[(diamond_table <lower_limit) | (diamond_table >upper_limit)]
8 table_outliers
```

Out[41]:

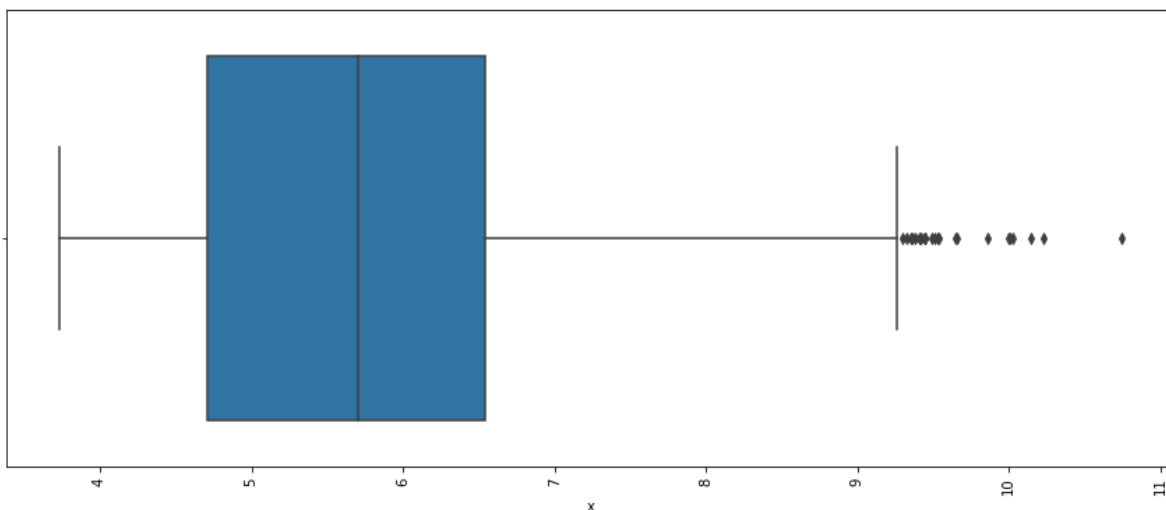
```
2          65.0
43          63.0
91          69.0
115         53.0
122         63.0
...
53825       63.0
53828       63.0
53840       63.0
53881       53.0
53897       63.0
Name: table, Length: 1966, dtype: float64
```

In [42]:

```
1 diamond_data = diamond_data[(diamond_data["table"]<80)&(diamond_data["table"]>40)]
```

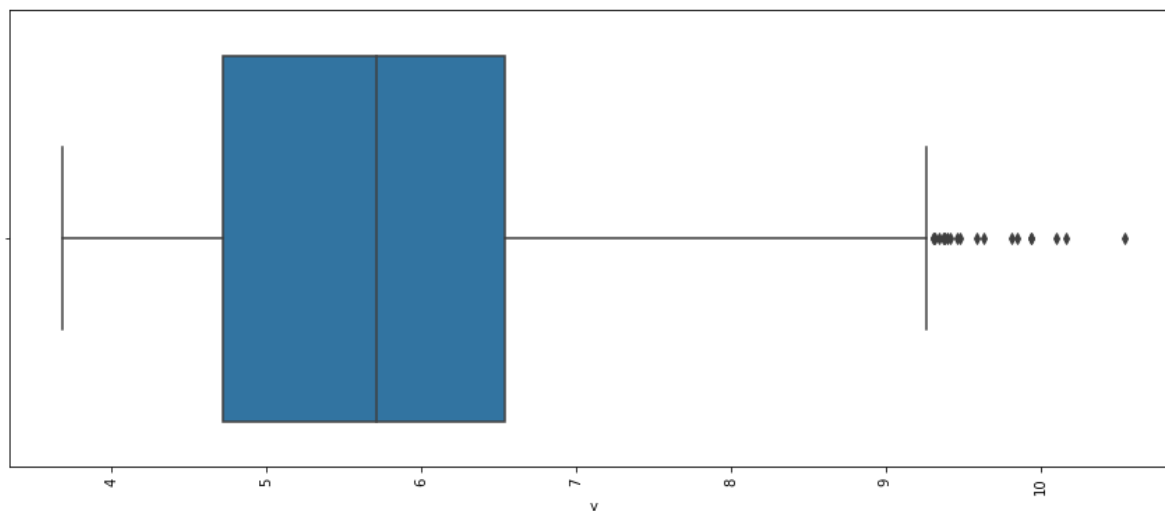
In [50]:

```
1 plt.figure(figsize=(15,6))
2 sns.boxplot(diamond_data['x'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



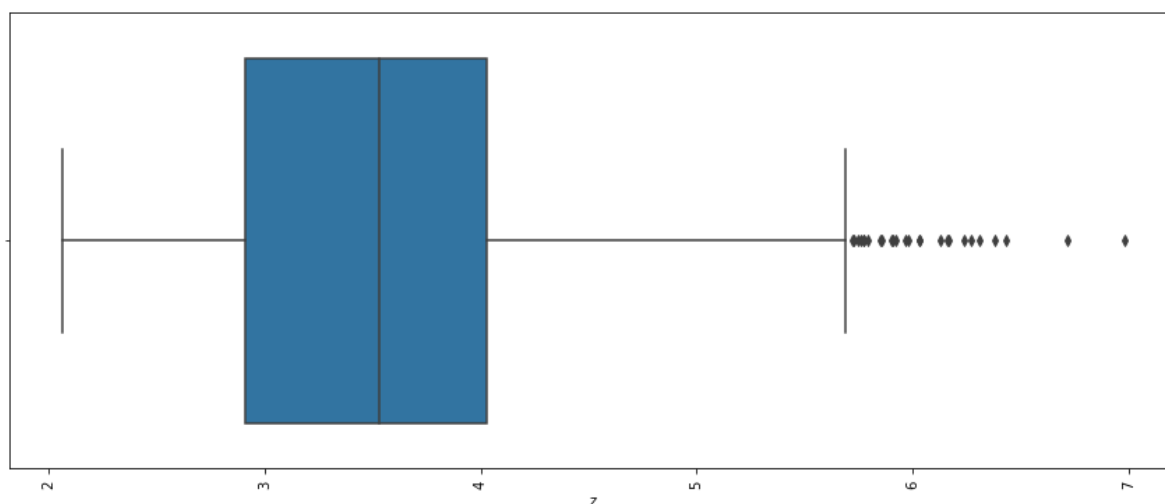
In [51]:

```
1 plt.figure(figsize=(15,6))
2 sns.boxplot(diamond_data['y'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [52]:

```
1 plt.figure(figsize=(15,6))
2 sns.boxplot(diamond_data['z'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [43]:

```
1 diamond_data = diamond_data[(diamond_data["x"]<30)]
2 diamond_data = diamond_data[(diamond_data["y"]<30)]
3 diamond_data = diamond_data[(diamond_data["z"]<30)&(diamond_data["z"]>2)]
```

In [44]:

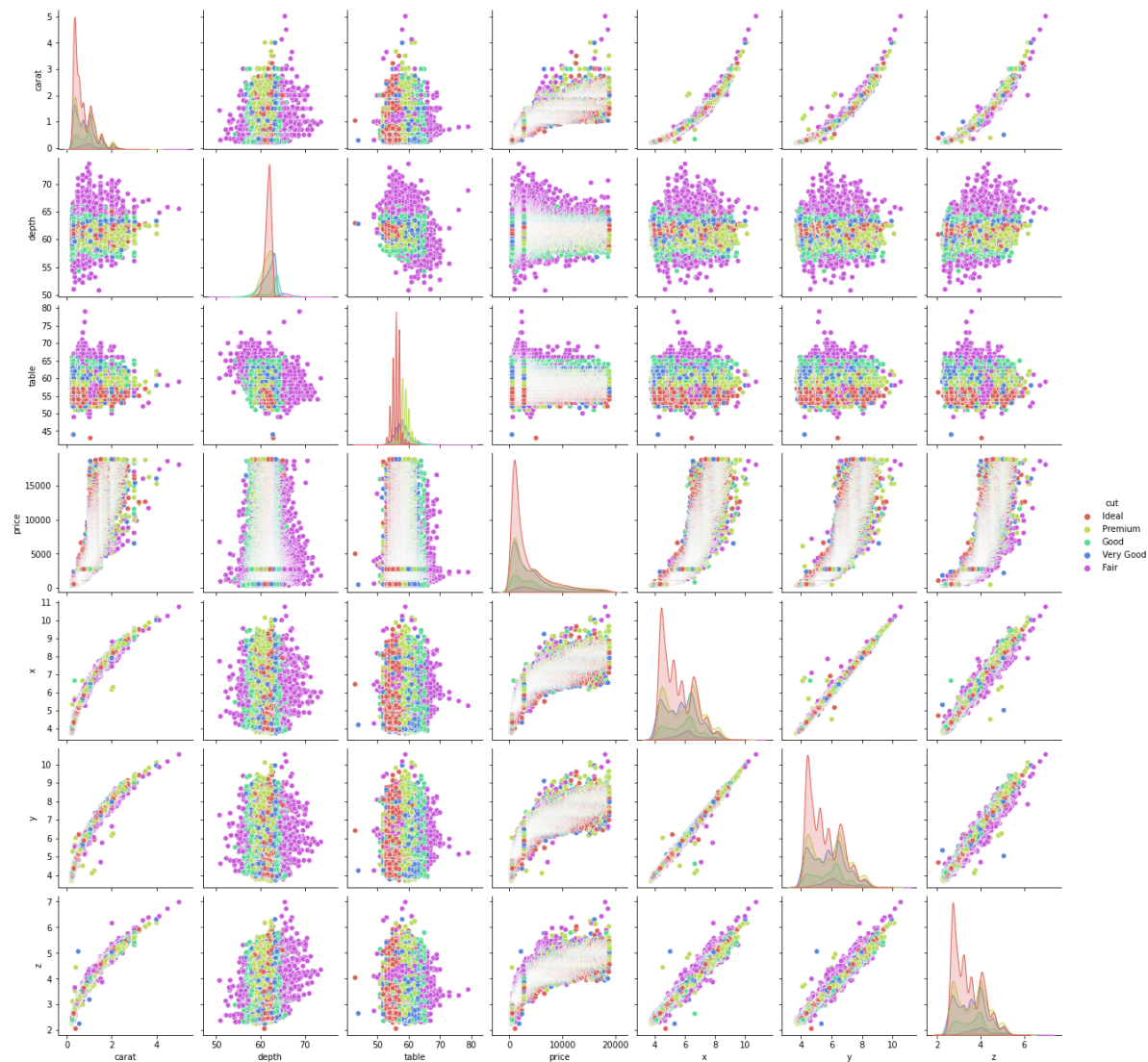
```
1 diamond_data.shape
```

Out[44]:

(53763, 10)

In [48]:

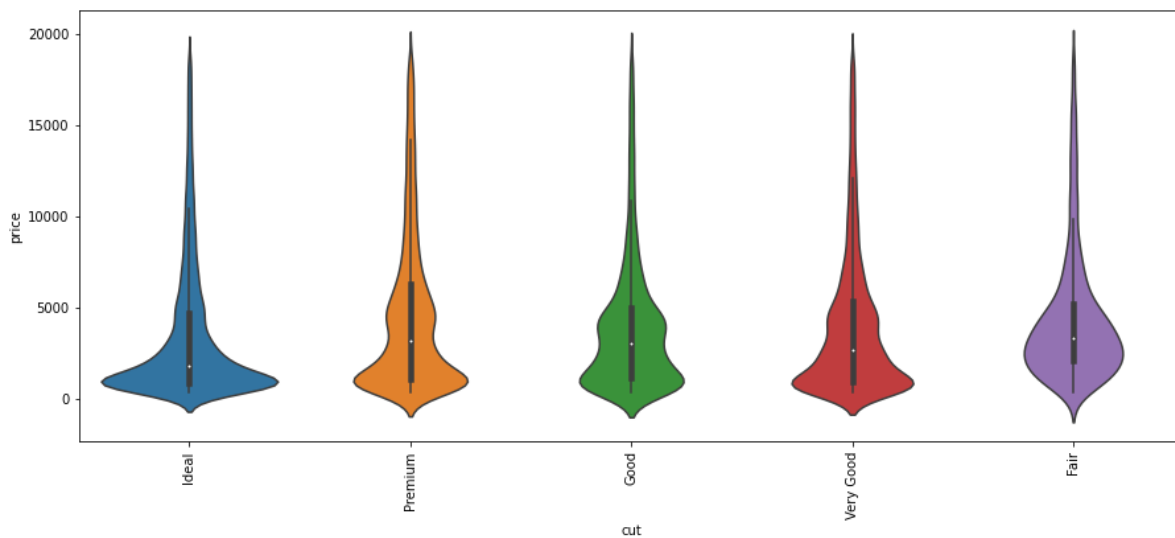
```
1 sns.pairplot(diamond_data, hue= "cut", palette = 'hls')  
2 plt.show()
```



In [54]:



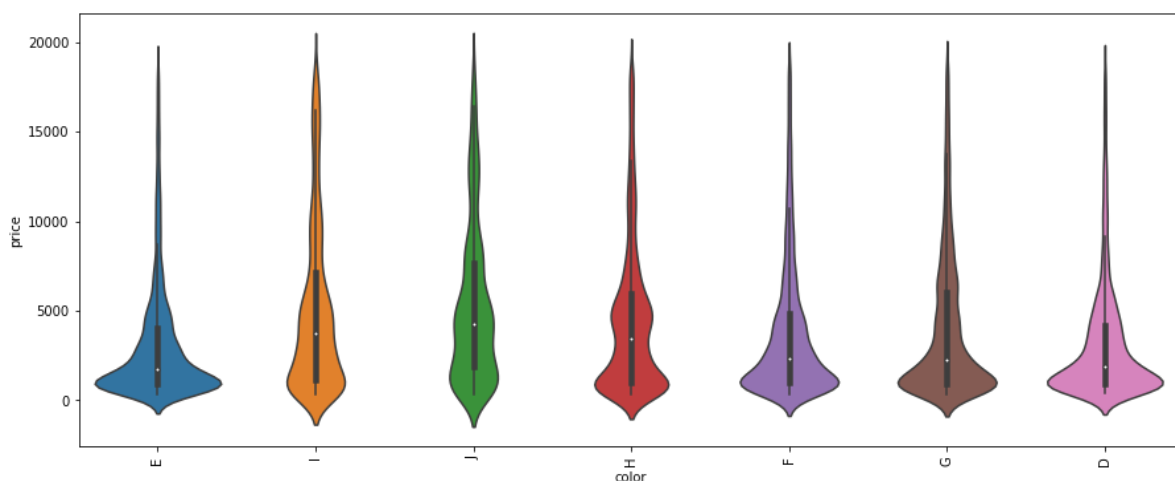
```
1 plt.figure(figsize=(15,6))
2 sns.violinplot(x = 'cut' , y = 'price' , data = diamond_data)
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [55]:

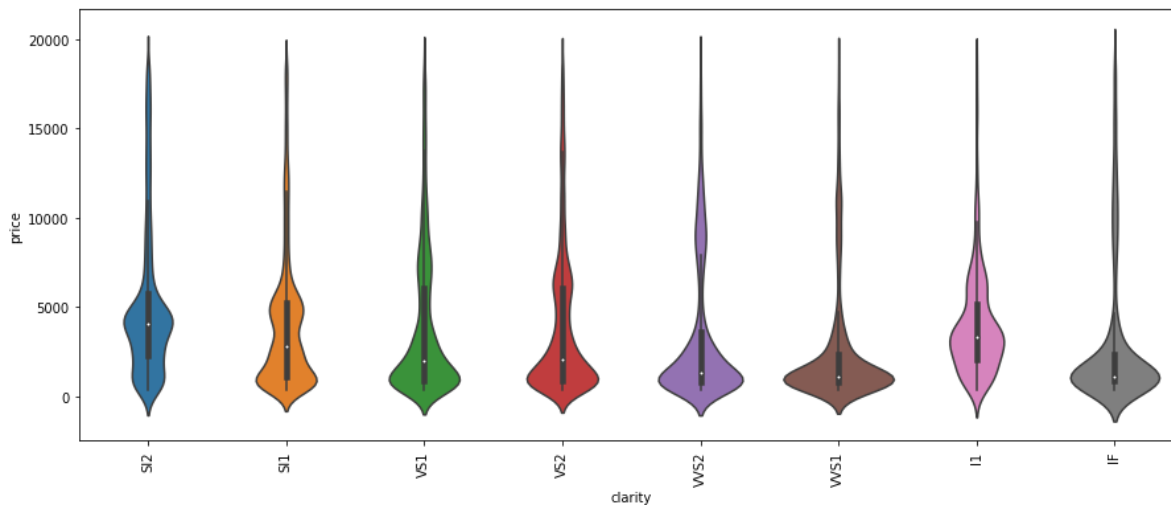


```
1 plt.figure(figsize=(15,6))
2 sns.violinplot(x = 'color' , y = 'price' , data = diamond_data)
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [56]:

```
1 plt.figure(figsize=(15,6))
2 sns.violinplot(x = 'clarity' , y = 'price' , data = diamond_data)
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [57]:

```
1 from sklearn.preprocessing import OneHotEncoder, LabelEncoder
2 from sklearn.model_selection import train_test_split
3 from sklearn.preprocessing import StandardScaler
4 from sklearn.decomposition import PCA
5 from sklearn.pipeline import Pipeline
6 from sklearn.tree import DecisionTreeRegressor
7 from sklearn.ensemble import RandomForestRegressor
8 from sklearn.linear_model import LinearRegression
9 from xgboost import XGBRegressor
10 from sklearn.neighbors import KNeighborsRegressor
11 from sklearn.model_selection import cross_val_score
12 from sklearn.metrics import mean_squared_error
13 from sklearn import metrics
```

In [59]:

```
1 s = (diamond_data.dtypes == "object")
2 object_cols = list(s[s].index)
```

In [60]:

```
1 label_data = diamond_data.copy()
2 label_encoder = LabelEncoder()
3 for col in object_cols:
4     label_data[col] = label_encoder.fit_transform(label_data[col])
```

In [61]:



```
1 label_data.head()
```

Out[61]:

	carat	cut	color	clarity	depth	table	price	x	y	z
0	0.23	2	1	3	61.5	55.0	326	3.95	3.98	2.43
1	0.21	3	1	2	59.8	61.0	326	3.89	3.84	2.31
2	0.23	1	1	4	56.9	65.0	327	4.05	4.07	2.31
3	0.29	3	5	5	62.4	58.0	334	4.20	4.23	2.63
4	0.31	1	6	3	63.3	58.0	335	4.34	4.35	2.75

In [62]:



```
1 diamond_data.describe()
```

Out[62]:

	carat	depth	table	price	x	y
count	53763.000000	53763.000000	53763.000000	53763.000000	53763.000000	53763.000000
mean	0.797460	61.748781	57.457207	3930.785336	5.731405	5.733299
std	0.473136	1.419309	2.226311	3985.807738	1.118563	1.110473
min	0.200000	50.800000	43.000000	326.000000	3.730000	3.680000
25%	0.400000	61.000000	56.000000	950.000000	4.710000	4.720000
50%	0.700000	61.800000	57.000000	2401.000000	5.700000	5.710000
75%	1.040000	62.500000	59.000000	5324.000000	6.540000	6.540000
max	5.010000	73.600000	79.000000	18823.000000	10.740000	10.540000



In [63]:

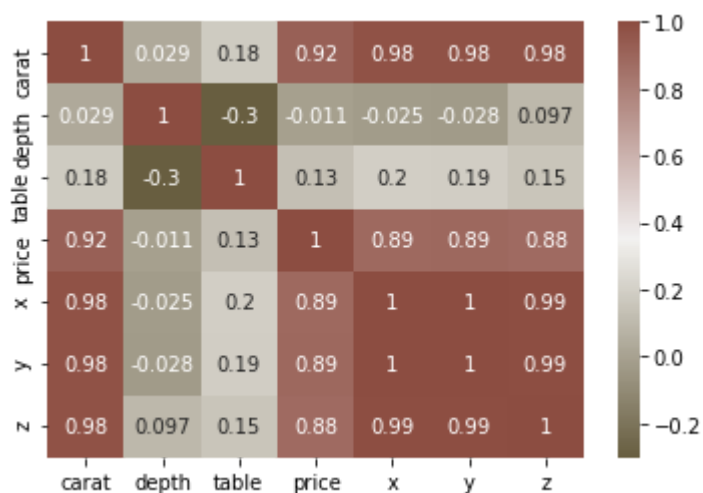
```
1 diamond_data.corr()
```

Out[63]:

	carat	depth	table	price	x	y	z
carat	1.000000	0.028718	0.181098	0.921546	0.977855	0.976940	0.977069
depth	0.028718	1.000000	-0.301988	-0.010766	-0.024679	-0.027810	0.097230
table	0.181098	-0.301988	1.000000	0.126425	0.195530	0.189325	0.154900
price	0.921546	-0.010766	0.126425	1.000000	0.887118	0.888708	0.882504
x	0.977855	-0.024679	0.195530	0.887118	1.000000	0.998654	0.991666
y	0.976940	-0.027810	0.189325	0.888708	0.998654	1.000000	0.991330
z	0.977069	0.097230	0.154900	0.882504	0.991666	0.991330	1.000000

In [65]:

```
1 cmap = sns.diverging_palette(70,20,s=50, l=40, n=6,as_cmap=True)
2 sns.heatmap(diamond_data.corr(), cmap=cmap,annot=True, )
3 plt.show()
```



In [66]:

```
1 X= label_data.drop(["price"],axis =1)
2 y= label_data["price"]
3 X_train, X_test, y_train, y_test = train_test_split(X, y,
4                                                     test_size=0.25,
5                                                     random_state=7)
```

In [67]:

```
1 pipeline_lr=Pipeline([("scalar1",StandardScaler()),
2                        ("lr_classifier",LinearRegression())])
3
4 pipeline_dt=Pipeline([("scalar2",StandardScaler()),
5                        ("dt_classifier",DecisionTreeRegressor())])
6
7 pipeline_rf=Pipeline([("scalar3",StandardScaler()),
8                        ("rf_classifier",RandomForestRegressor())])
9
10 pipeline_kn=Pipeline([("scalar4",StandardScaler()),
11                       ("rf_classifier",KNeighborsRegressor())])
12
13 pipeline_xgb=Pipeline([("scalar5",StandardScaler()),
14                        ("rf_classifier",XGBRegressor())])
15
16 pipelines = [pipeline_lr, pipeline_dt, pipeline_rf, pipeline_kn,
17              pipeline_xgb]
18 pipe_dict = {0: "LinearRegression", 1: "DecisionTree", 2: "RandomForest",3: "KNeigh
19
20 for pipe in pipelines:
21     pipe.fit(X_train, y_train)
```

In [68]:

```
1 cv_results_rms = []
2 for i, model in enumerate(pipelines):
3     cv_score = cross_val_score(model, X_train,y_train,scoring="neg_root_mean_square
4     cv_results_rms.append(cv_score)
5     print("%s: %f " % (pipe_dict[i], cv_score.mean()))
```

LinearRegression: -1333.321776

DecisionTree: -754.017952

RandomForest: -551.247491

KNeighbors: -828.946820

XGBRegressor: -551.226106

In [71]:

```
1 pred = pipeline_xgb.predict(X_test)
```

In [70]:

```
1 print("Training Accuracy :", pipeline_xgb.score(X_train, y_train))
2 print("Testing Accuracy :", pipeline_xgb.score(X_test, y_test))
```

Training Accuracy : 0.9911507531791393

Testing Accuracy : 0.9812464134865915

In [72]:



```
1 print("R^2:",metrics.r2_score(y_test, pred))
2 print("Adjusted R^2:",1 - (1-metrics.r2_score(y_test, pred))*(len(y_test)-1)/(len(y.
3 print("MAE:",metrics.mean_absolute_error(y_test, pred))
4 print("MSE:",metrics.mean_squared_error(y_test, pred))
5 print("RMSE:",np.sqrt(metrics.mean_squared_error(y_test, pred)))
```

R^2: 0.9812464134865915
Adjusted R^2: 0.9812338468661894
MAE: 281.2550541364262
MSE: 294422.2563174091
RMSE: 542.6069077310103