#Program 1

Plot a boxplot for price over Tuterxfrom the dataset diamond.csverx Which of the categories under Cuterxhave the highest median price?

```
import pandas as pd
import matplotlib.pyplot as plt

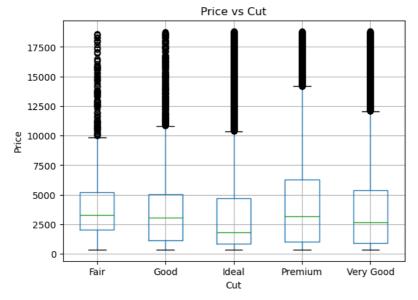
# Read the dataset
df = pd.read_csv("diamonds.csv")

print(df.head())

# Create the boxplot
plt.figure(figsize=(8, 6))
df.boxplot(column='price', by='cut')
plt.xlabel('Cut')
plt.ylabel('Price')
plt.title('Price vs Cut')
plt.suptitle('') # Remove the default title
plt.show()
```

```
z
0 2.43
1 2.31
2 2.31
3 2.63
4 2.75
```

<Figure size 800x600 with 0 Axes>



#Program 2

'''Create a frequency table (one-way table) for the variable $\overline{\mathtt{Cut}}$ from the dataset $\overline{\mathtt{diamond.csv}}$ What is the frequency for the cut type $\overline{\mathtt{Idea}}$!''

```
import pandas as pd

# Read the dataset
df = pd.read_csv("diamonds.csv")

# Create the frequency table
frequency_table = df['cut'].value_counts()

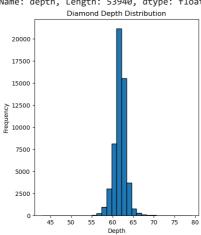
# Print the frequency table
print(frequency_table)

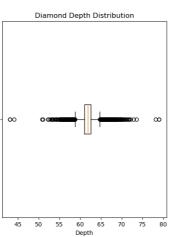
# Access the frequency for the cut type "Ideal"
ideal_frequency = frequency_table['Ideal']
```

plt.show()

```
print("Frequency for cut type 'Ideal':", ideal_frequency)
                   21551
     Ideal
                   13791
     Premium
     Very Good
                   12082
     Good
                    4906
     Fair
                    1610
     Name: cut, dtype: int64
     Frequency for cut type 'Ideal': 21551
#Program 3
# Show the subplot of the diamond carat weight distribution.
import pandas as pd
import matplotlib.pyplot as plt
# Read the dataset
df = pd.read_csv("diamonds.csv")
print(df['carat'])
# Create a subplot with 1 row and 2 columns
fig, axs = plt.subplots(1, 2, figsize=(12, 6))
# Plot the histogram of carat weight on the first subplot
axs[0].hist(df['carat'], bins=30, edgecolor='black')
axs[0].set_xlabel('Carat Weight')
axs[0].set_ylabel('Frequency')
axs[0].set_title('Carat Weight Distribution')
# Plot the boxplot of carat weight on the second subplot
axs[1].boxplot(df['carat'], vert=False)
axs[1].set_yticklabels('')
axs[1].set_xlabel('Carat Weight')
axs[1].set_title('Carat Weight Distribution')
# Adjust the spacing between subplots
plt.subplots_adjust(wspace=0.3)
# Show the plot
```

```
0.23
              0.21
              0.23
#Program 4
# Show the subplot of diamond depth distribution.
import pandas as pd
import matplotlib.pyplot as plt
# Read the dataset
df = pd.read_csv("diamonds.csv")
print(df['depth'])
# Create a subplot with 1 row and 2 columns
fig, axs = plt.subplots(1, 2, figsize=(12, 6))
\ensuremath{\mathtt{\#}} Plot the histogram of diamond depth on the first subplot
axs[0].hist(df['depth'], bins=30, edgecolor='black')
axs[0].set_xlabel('Depth')
axs[0].set_ylabel('Frequency')
axs[0].set_title('Diamond Depth Distribution')
# Plot the boxplot of diamond depth on the second subplot
axs[1].boxplot(df['depth'], vert=False)
axs[1].set_yticklabels('')
axs[1].set_xlabel('Depth')
axs[1].set_title('Diamond Depth Distribution')
# Adjust the spacing between subplots
plt.subplots_adjust(wspace=0.4)
# Show the plot
plt.show()
     0
              61.5
     1
              59.8
     2
              56.9
              62.4
     4
              63.3
     53935
              60.8
     53936
              63.1
     53937
              62.8
     53938
              61.0
     53939
              62.2
     Name: depth, Length: 53940, dtype: float64
                   Diamond Depth Distribution
```





```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
```

```
# Read the dataset
df = pd.read_csv("diamonds.csv")
# Define the input features (X) and target variable (y)
X = df[['carat', 'cut', 'clarity', 'depth']]
y = df['price']
# Convert categorical variables to numerical using one-hot encoding
X_encoded = pd.get_dummies(X)
print("X_encoded=",X_encoded)
\ensuremath{\text{\#}} Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_encoded, y, test_size=0.2, random_state=42)
# Build the linear regression model
model = LinearRegression()
model.fit(X_train, y_train)
# Make predictions on the testing set
y_pred = model.predict(X_test)
# Calculate the accuracy (R-squared score)
accuracy = r2_score(y_test, y_pred)
print("Accuracy:", accuracy)
                       carat depth cut_Fair cut_Good cut_Ideal cut_Premium \
     X encoded=
             0.23 61.5
                                 0
                                            0
                                                       1
             0.21
                    59.8
                                 0
                                            a
                                                       a
     1
                                                                    1
     2
             0.23
                    56.9
                                 0
                                            1
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                                                                    0
     3
             0.29
                    62.4
                                 0
                                            0
                                                       a
                                                                    1
     4
             0.31
                    63.3
                                 0
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     53935
             0.72
                    60.8
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     53936
             0.72
                    63.1
                                            1
     53937
             0.70
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                    62.8
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                                                       0
     53938
             0.86
                    61.0
                                            0
                                 0
                                                       0
                                                                    1
     53939
             0.75
                                            0
                                 0
                   62.2
            cut_Very Good clarity_I1 clarity_IF clarity_SI1 clarity_SI2 \
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     1
                        a
                                     a
                                                 a
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     2
                        a
                                     0
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     3
                        0
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     4
                        0
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     53936
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                                                              a
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            clarity_VS1 clarity_VVS2 clarity_VVS1 clarity_VVS2
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     53937
                      0
                                    0
                                                  a
                                                                0
     53938
                      0
                                    0
                                                  a
                                                                0
     53939
                      0
     [53940 rows x 15 columns]
```

LinearRegression() Accuracy: 0.8966651780887069