Data Visualization Lab 2

Attribute Dataset

Utkarsh Bhangale 20200802124

Exploratory Data Analysis (EDA)

load the data and perform some initial analysis. Assuming you have the dataset in a CSV file:

```
In [1]:
```

```
# Import necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
# Load the dataset
data = pd.read csv('Attribute dataset.csv')
# Check the first few rows of the dataset
print(data.head())
# Summary statistics
print(data.describe())
# Data types and missing values
print(data.info())
Dress_ID Style Price Rating Size Season NeckLine SleeveLength \
0 1006032852 Sexy Low 4.6 M Summer o-neck sleevless
1 1212192089 Casual Low 0.0 L Summer o-neck Petal
2 1190380701 vintage High 0.0 L Automn o-neck full
3 966005983 Brief Average 4.6 L Spring o-neck full
4 876339541 cute Low 4.5 M Summer o-neck butterfly
          Material FabricType Decoration Pattern Type Recommendation
           NaN chiffon ruffles animal
0
      microfiber NaN ruffles
polyster NaN NaN
1
     microfiber
                                                           animal
                                                                                         0
                                                           print
2
                                                                                         0
3 silk chiffon embroidary print
4 chiffonfabric chiffon bow dot
             Dress ID Rating Recommendation
                                              0.427975
count 4.790000e+02 479.000000 479.000000
mean 9.022420e+08 3.489353
std 1.822352e+08 2.028941
      0.000000
min
                                                  0.000000
25%
                                                  0.000000
50%
                             4.800000
                                                  1.000000
75%
max
                                                   1.000000
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 479 entries, 0 to 478
Data columns (total 13 columns):
 # Column Non-Null Count Dtype
                           _____
   Dress_ID 479 non-null int64
Style 479 non-null object
Price 477 non-null object
Rating 479 non-null float64
Size 479 non-null object
Season 477 non-null object
NeckLine 476 non-null object
SleeveLength 477 non-null object
Material 360 non-null object
FabricType 223 non-null object
 0 Dress_ID
 1 Style
 2 Price
 3 Rating
 5 Season
 7
 8
```

```
ranticilhe
                     ZZJ IIVII IIUTT
                                      UNJEUL
 10
    Decoration
                     255 non-null
                                      object
     Pattern Type
                     377 non-null
 11
                                      object
 12 Recommendation 479 non-null
                                      int64
dtypes: float64(1), int64(2), object(10)
memory usage: 48.8+ KB
None
```

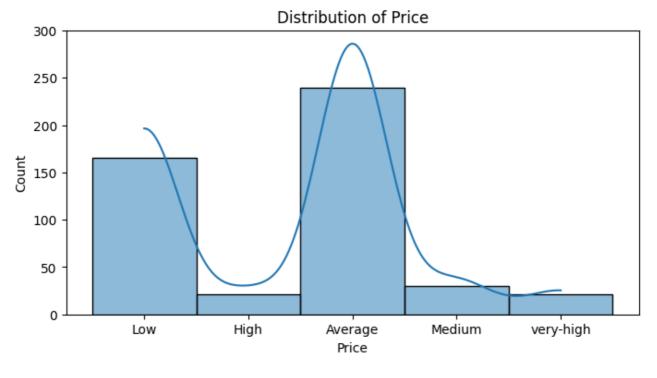
Univariate Analysis

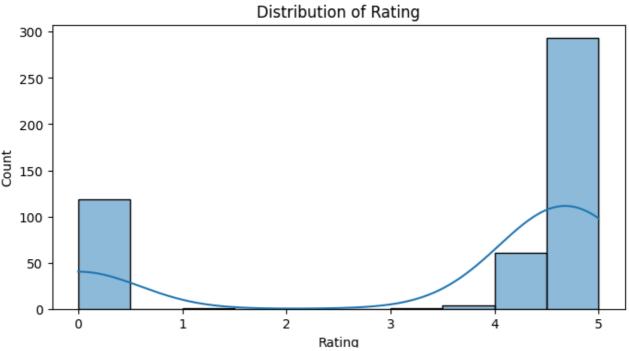
For univariate analysis, let's analyze individual variables in more detail. Here's an example of how you can create histograms and other visualizations for each numerical attribute:

In [2]:

```
# Univariate Analysis for numerical attributes
num_attributes = ['Price', 'Rating']

for attribute in num_attributes:
   plt.figure(figsize=(8, 4))
   sns.histplot(data=data, x=attribute, kde=True)
   plt.title(f'Distribution of {attribute}')
   plt.show()
```





For categorical attributes, you can create bar plots to visualize the distribution:

In [3]:

```
# Univariate Analysis for categorical attributes
cat_attributes = ['Style', 'Size', 'Season', 'NeckLine', 'SleeveLength', 'Material', 'Fa
bricType', 'Decoration', 'Pattern Type']

for attribute in cat_attributes:
   plt.figure(figsize=(10, 5))
   sns.countplot(data=data, x=attribute, order=data[attribute].value_counts().index)
   plt.xticks(rotation=45)
   plt.title(f'Count of {attribute}')
   plt.show()
```

