Data Wrangling Class Project

Vicky, 11/28/2023

Project Overview

The AuctionDataAnalysis class is tailored to dissect auction records of artworks cataloged in a CSV file. It executes an array of data preprocessing operations and generates visual representations to facilitate a comprehensive understanding of the dataset.

Class Methods:

- 1. __init__(file_path) : Initializes the class with the specified dataset file path 2. preprocess data(): Preprocesses the dataset, including standardizing price formats and organizing movement orders
- 3. get_author_name(): Retrieves the name of the report's author.
 4. get_record_num(): Provides the count of auction records in the dataset.
- 5. get_top_artists(): Presents the total count of unique artists and lists the top ten contributing artists
- 6. get_top_titles(): Displays the count of unique titles and highlights the most frequent five titles in the dataset.
- 7. visualize_period_statistics(): Generates a table detailing the Combined Price Statistics for Each Period (sorted in descending order by counts) and provides visual insights into Mean Price and Counts for Each Period.

 8. visualize_movement_statistics(): Produces a table outlining the Combined Price Statistics for Each Movement (sorted in ascending order based on the historical timeline of movements) and showcases Mean Price and Counts for Each Movement.

Additional Information:

The file includes an appendix explaining the Historical Timeline of Movements.

Dataset Information:

The class is tested using the Sotheby's art price dataset from Kaggle.

Citation: Fl.Kuhm. (2022). Art Price Dataset [Dataset]. Kaggle. DOI

The dataset comprises artworks and sculptures listed for sale on Sothebys, encompassing artist names, prices, associated time periods, and art movements

Class AuctionDataAnalysis:

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

```
class AuctionDataAnalysis:
    def __init__(self, file_path):
        self.data = pd.read_csv(file_path)
        self.processed_data = None
        self.record_num = None
        self.author = 'Vicky'
In [10]:
                              def preprocess_data(self):
    print("\033[1mSummary of the Dataset\033[0m\n")
                                     self.processed_data = self.data.iloc[:, [1, 2, 3, 4, 7, 8]]
self.processed_data.loc[:, 'price'] = self.processed_data['price'].str.split().str[0]
self.processed_data.loc[:, 'price'] = self.processed_data['price'].astype(float)
self.processed_data.loc[self.processed_data['price'] < 180, 'price'] = 1800
self.processed_data.dropna(subset['qriist'], inplace=True)
                                      self.processed_data.loc[:, 'movement'] = pd.Categorical(
    self.processed_data['movement'],
    categories=movement_order,
    ordered=frue
                                       self.record_num = self.record_num = len(self.processed_data)
                              def get_author_name(self):
    return self.author
                              def get_record_num(self):
    return self.record_num
                              def get_top_artists(self):
                                      yex_vop_artistiseu;; artistiseu;; artisti].nunique() artist_nun self_processed_data['artist'].nunique() print(f"2. There are records attributed to {artist_num} different artists. Notably, among these, the following ten artists have contributed the most artworks:\n") artist_counts = self_processed_data['artist'].value_counts() return artist_counts.lead(18)
                                      ye__top_tittes(set7):
title_num = self.processed_data['title'].nunique()
print("3. This dataset contains {title_num} unique titles. Among these, the top five titles have the highest frequency of artworks:\n")
title_counts = self.processed_data['title'].value_counts()
return title_counts.head(counts.head()).
                              def get top titles(self):
```

```
def visualize_period_statistics(self):
                  if self.processed_data is None:
    print("No data available. Preprocess the data first.")
    return
               period_counts = self.processed_data['period'].value_counts()
mean_price_period = self.processed_data.groupby('period')['price'].mean().round(2)
median_price_period = self.processed_data.groupby('period')['price'].max()
max_price_period = self.processed_data.groupby('period')['price'].mix()
sd_price_period = self.processed_data.groupby('period')['price'].mix()
sd_price_period = self.processed_data.groupby('period')['price'].mix()
                combined_stats_period = pd.concat([period_counts, mean_price_period, median_price_period, max_price_period, min_price_period, sd_price_period], axis=1) combined_stats_period.columns = ['Counts', 'Mean Price', 'Mean Price', 'Max Price', 'Min Price', 'Standard Deviation']
                pd.set_option('display.width', 1000)
pd.set_option('display.max_colwidth', None)
                \label{print print pri
                period_mean_price = combined_stats_period['Mean Price'].sort_values()
period_counts = combined_stats_period['Counts'].reindex(period_mean_price.index)
                fig, ax1 = plt.subplots(figsize=(10, 6))
                ax1.bar(range(len(period_mean_price)), period_mean_price, color='tan', label='Mean Price')
ax1.set_xlabel('Period')
ax1.set_ylabel('Mean Price', color='sienna')
ax1.tick_params(axis='y', labelcolor='sienna')
                ax2 = ax1.twinx()
ax2.plot(range(len(period_counts)), period_counts, marker='o', linestyle='-', color='brown', label='Counts')
ax2.set_vlabel('Counts', color='brown')
ax2.tick_params(axis='y', labelcolor='brown')
                plt.title('Mean Price and Counts for Each Period')
ax1.legend(loc='upper left')
ax2.legend(loc='upper right')
                ax1.set_xticks(range(len(period_mean_price)))
ax1.set_xticklabels(period_mean_price.index, rotation=45, ha='right')
                plt.show()
```

```
def visualize_novement_statistics(self):
    if self.processed_data is Nome:
        print("No data available. Proprocess the data first.")
        return
    novement_counts = self.processed_data("novement').value_counts()
    mean_price_novement = self.processed_data(groupby("novement')['price'].mean().
    mean_price_novement = self.processed_data(groupby("novement')['price'].mean().
    init_price_novement = self.processed_data(groupby("novement')['price'].mean().
    init_price_novement = self.processed_data(groupby("novement')['price'].self.price().
    init_price_novement = self.processed_data(groupby("novement')['price'].self.price().
    combined_stata_novement = nd.concat(!novement_counts, mean_price_novement, mean_price_novement, nd.price_novement, nd.price_novement
```

Test:

```
In [11]:
    file_path = 'ClassProj/artDataset.csv'
    art_analysis = AuctionDataAnalysis(file_path)
    art_analysis.perprocess_data()

author_name = art_analysis.get_author_name()
    print(""Authored by {author_name\n")

    record_num = art_analysis.get_record_num()
    print(""I. The dataset comprises {record_num\ auction records of artworks.\n")

    top_artists = art_analysis.get_top_artists()
    print("'(top_artists)\n')
    top_titles = art_analysis.get_top_titles()
    print("'(top_titles)\n')
    art_analysis.visualize_period_statistics()

art_analysis.visualize_novement_statistics()
```

Summary of the Dataset

Authored by Vicky

- 1. The dataset comprises 753 auction records of artworks.
- 2. There are records attributed to 454 different artists. Notably, among these, the following ten artists have contributed the most artworks:

Russell Young 17
John Fischer 15
Ruth Bernhard 13
Donald Sultan 9
Grant Hacking 8
Richard Bernstein 8
Ed Ruscha 8
Cindy Sherman 7
Robert Indiana 7
Kim Gottlieb Walker 7
Name: artist, dtype: int64

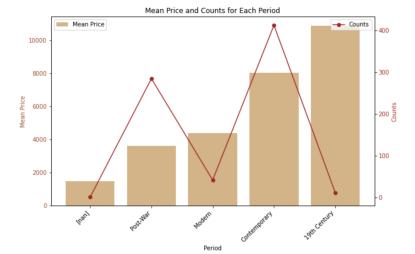
3. This dataset contains 678 unique titles. Among these, the top five titles have the highest frequency of artworks:

Untitled 35
Untitled 10
Madame de Pompadour (née Poisson) 7
Rays 5
Fragments Of Hope 5
Name: title, dtype: int64

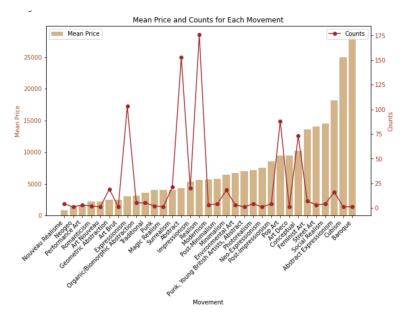
Name: title, dtype: int64

4. Combined Price Statistics for Each Period: (sorted in descending order by counts)

	Counts	Mean Price	Median Price	Max Price	Min Price	Standard Deviation
Contemporary	413	8060.65	3995.0	75000.0	128.0	10901.45
Post-War	285	3608.21	1500.0	70000.0	595.0	7157.71
Modern	42	4382.86	2775.0	25000.0	600.0	4672.35
19th Century	12	10900.00	8575.0	34000.0	2250.0	9519.48
[nan]	1	1500.00	1500.0	1500.0	1500.0	NaN



5. Combined Price Statistics for Each	Movement	: (sorted in	ascending ord	ler based on	the histor	ical timeline of movements)
	Counts	Mean Price	Median Price	Max Price	Min Price	Standard Deviation
Baroque	1	28500.00	28500.0	28500.0	28500.0	NaN
Romanticism	2	2250.00	2250.0	2500.0	2000.0	353.55
Realism	176	5647.19	3000.0	50000.0	600.0	7802.75
Impressionism	20	5306.50	2027.5	37000.0	800.0	8554.51
Traditional	5	3620.00	1500.0	12000.0	800.0	4769.91
Post-Impressionism	4	8500.00	6750.0	15500.0	5000.0	4949.75
Art Nouveau	1	2250.00	2250.0	2250.0	2250.0	NaN
Expressionism	103	3064.37	1500.0	40000.0	800.0	4982.24
Cubism	1	25000.00	25000.0	25000.0	25000.0	NaN
Abstract	153	4303.89	1275.0	45000.0	595.0	8135.41
Modernism	3	5700.00	3550.0	10000.0	3550.0	3723.91
Geometric Abstraction	19	2449.21	1275.0	15000.0	680.0	3191.94
Surrealism	21	4046.19	1500.0	24000.0	800.0	6256.84
Art Deco	1	9500.00	9500.0	9500.0	9500.0	NaN
Magic Realism	1	4000.00	4000.0	4000.0	4000.0	NaN
Organic/Biomorphic Abstraction	5	3163.00	680.0	12500.0	680.0	5225.90
Social Realism	4	14500.00	16500.0	18000.0	7000.0	5196.15
Abstract Expressionism	16	18195.94	9287.5	70000.0	1200.0	20211.27
Art Brut	1	2500.00	2500.0	2500.0	2500.0	NaN
Pop Art	88	9477.70	3947.5	75000.0	128.0	13280.53
Minimalism	18	6457.22	1500.0	71500.0	680.0	16409.77
Nouveau Réalisme	4	850.00	850.0	900.0	800.0	57.74
Environmental Art	3	6666.67	8000.0	8000.0	4000.0	2309.40
Street Art	3	14000.00	14000.0	14000.0	14000.0	0.00
Conceptual	73	10230.14	7500.0	48000.0	800.0	10510.10
Feminist Art	7	13571.43	12000.0	25000.0	6000.0	8541.22
Photorealism	4	7158.75	9500.0	9500.0	135.0	4682.50
Post-Minimalism	4	5750.00	5500.0	8000.0	4000.0	2061.55
Performance Art	3	1666.67	2000.0	2000.0	1000.0	577.35
Neo-Expressionism	1	7550.00	7550.0	7550.0	7550.0	NaN
Punk, Young British Artists, Abstract	1	7000.00	7000.0	7000.0	7000.0	NaN
Punk	2	4000.00	4000.0	4000.0	4000.0	0.00
Neogeo	1	1500.00	1500.0	1500.0	1500.0	NaN



Appendix1: Historical Timeline of Movements

- 1. Baroque: 17th first half of the 18th century
- 2. Romanticism: 1805 1830s
- 3. Realism: 1840 the late nineteenth century
- 4. Impressionism: 1867 1886
- 5. Traditional: 1880 1900
- 6. Post-Impressionism: 1886 1905
- 7. Art Nouveau: 1890 1910
- 8. Expressionism: approximately 1905 1920
- 9. Cubism: 1907 1914
- 10. Abstract: early 20th century
- 11. Modernism: early 1900s the early 1940s
- 12. Geometric Abstraction: 1910 1960
- 13. Surrealism: late 1910s early 1920s
- 14. Art Deco: 1919 1939
- 15. Magic Realism: 1920s 1940s
- 16. Organic/Biomorphic Abstraction: 1930s/40s
- 17. Social Realism: 1930s
- 18. Abstract Expressionism: 1943 the mid-1950s
- 19. Art Brut: 1945
- 20. Pop Art: 1950s 1960s
- 21. Minimalism: 1960s 1970s
- 22. Nouveau Réalisme: 1960s
- 23. Environmental Art: 1960s
- 24. Street Art: 1960s 1980s
- 25. Conceptual: mid-1960s the mid-1970s
- 26. Feminist Art: late 1960s
- 27. Photorealism: late 1960s
- 28. Post-Minimalism: late 1960s early 1970s
- 29. Performance Art: 1970s
- 30. Neo-Expressionism: late 1970s and 1980s 31. Punk, Young British Artists, Abstract: 1975 – 1980
- 32. Punk: 1975 1980
- 33. Neogeo: 1990s
- 34. [nan]