

annand_python_project

August 25, 2023

1 DSE5002 Python Project

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1.2 8/25/2023

1.2.1 Import Libraries

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

1.2.2 Import Data Sets

```
[2]: cost_of_living_data = pd.read_csv("cost_of_living.csv")
ds_salaries = pd.read_csv("ds_salaries.csv")
levels_fyi_salary_data = pd.read_csv("Levels_Fyi_Salary_Data.csv")
country_codes_data = pd.read_excel("country_codes.xlsx")
```

1.2.3 Wrangle Levels.fyi Data

```
[3]: levels_fyi_data = pd.DataFrame(levels_fyi_salary_data.
    ↳loc[levels_fyi_salary_data['title'].str.contains("(D|d)ata"), :])
new = levels_fyi_data["location"].str.split(", ", n = -1, expand = True)
levels_fyi_data['city'] = new[0]
levels_fyi_data['state.or.province'] = new[1]
levels_fyi_data['country'] = new[2]
levels_fyi_data.rename(columns={'title': 'job_title', 'totalyearlycompensation': '
    ↳salary_in_usd'}, inplace=True)
levels_fyi_data.reset_index(inplace=True)
for i in range(len(levels_fyi_data)):
    if pd.isnull(levels_fyi_data.at[i, 'country']):
        levels_fyi_data.loc[i, 'country'] = "United States"
```

C:\Users\janna\AppData\Local\Temp\ipykernel_23848\4087661922.py:1: UserWarning:
This pattern is interpreted as a regular expression, and has match groups. To
actually get the groups, use str.extract.

```
levels_fyi_data = pd.DataFrame(levels_fyi_salary_data.loc[levels_fyi_salary_data['title'].str.contains("(D|d)ata"), :])
```

1.2.4 Wrangle ds_salaries Data

```
[4]: ds_salaries.rename(columns={'company_location': 'country'}, inplace=True)
for i in range(len(ds_salaries)):
    code = ds_salaries.at[i, 'country']
    conversion = pd.DataFrame(country_codes_data.loc[country_codes_data['Alpha-2_
↪code'] == code, 'Country']).iat[0,0]
    ds_salaries.loc[i, 'country'] = conversion
```

1.2.5 Wrangle cost_of_living Data

```
[5]: cost_of_living_data.rename(columns={'City': 'location'}, inplace=True)
new = cost_of_living_data['location'].str.split(", ", n=-1, expand=True)
cost_of_living_data['city'] = new[0]
cost_of_living_data['state.or.province'] = new[1]
cost_of_living_data['country'] = new[2]
for i in range(len(cost_of_living_data)):
    if pd.isnull(cost_of_living_data.at[i, 'country']):
        cost_of_living_data.loc[i, 'country'] = cost_of_living_data.at[i, 'state.
↪or.province']
```

1.2.6 Adjust country_codes data

```
[6]: us_code_index = country_codes_data.index[country_codes_data['Alpha-2 code'] ==
↪"US"].tolist()
country_codes_data.loc[us_code_index[0], 'Country'] = "United States"
```

1.2.7 Merge and subset salary data

```
[7]: all_salaries_df = pd.merge(levels_fyi_data,
                                ds_salaries,
                                how='outer')
cleaned_salary = all_salaries_df.loc[:,
                                ['job_title', 'salary_in_usd', 'city',
↪'state.or.province', 'country']]
```

1.2.8 Determine ratings by city

```
[8]: city_median_df = pd.DataFrame(cleaned_salary.groupby(by="city")['salary_in_usd'].
↪median())
city_median_df.reset_index(inplace=True)

city_rankings_df = pd.merge(city_median_df,
                             cost_of_living_data,
```

```

                                how='outer')
city_rankings_df = city_rankings_df.assign(cost_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Cost of Living Index'],
                                rent_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Rent Index'],
                                living_rent_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Cost of Living Plus Rent
↳Index'],
                                groceries_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Groceries Index'],
                                restaurant_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Restaurant Price Index'],
                                purchase_rating = lambda x:
↳city_rankings_df['salary_in_usd'] / city_rankings_df['Local Purchasing Power
↳Index'])

```

1.2.9 Visualizations for City Data

```

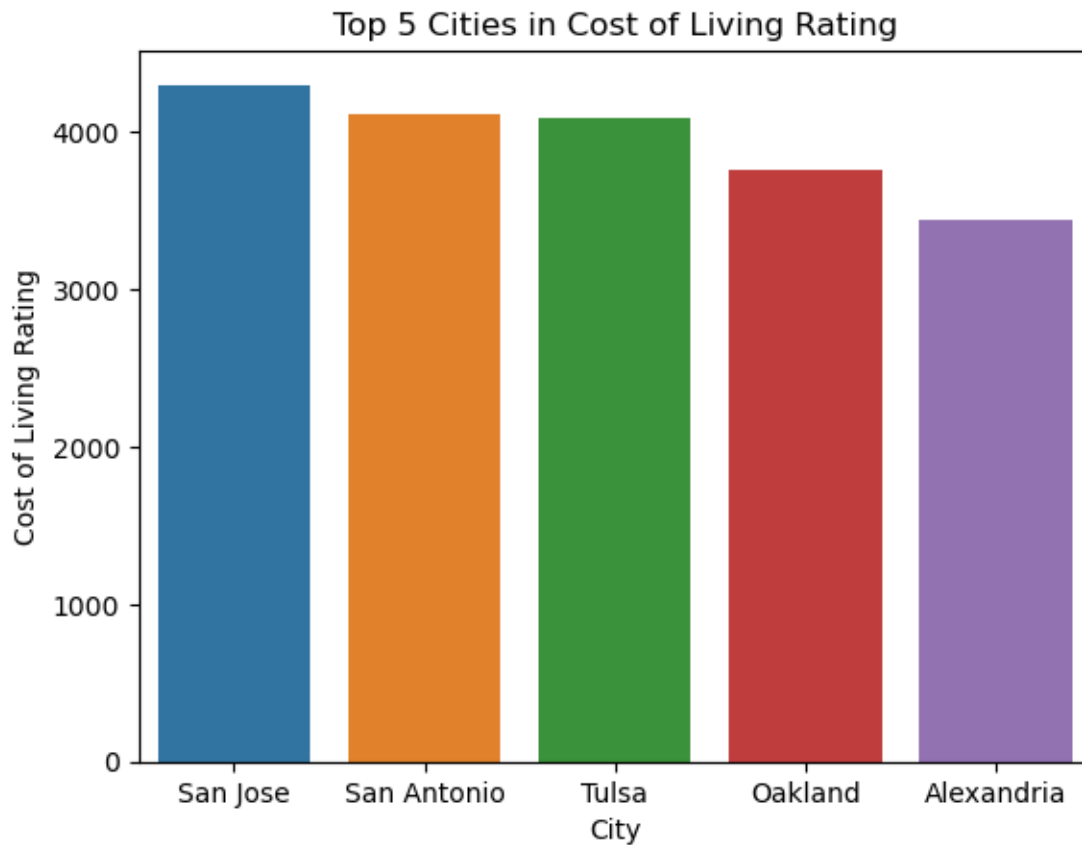
[9]: cost = city_rankings_df.sort_values(by=['cost_rating'], ascending=False).
↳reset_index().loc[0:4, ['city', 'cost_rating']]
sns.barplot(data=cost, x='city', y='cost_rating')
plt.xlabel("City")
plt.ylabel("Cost of Living Rating")
plt.title("Top 5 Cities in Cost of Living Rating")

```

```

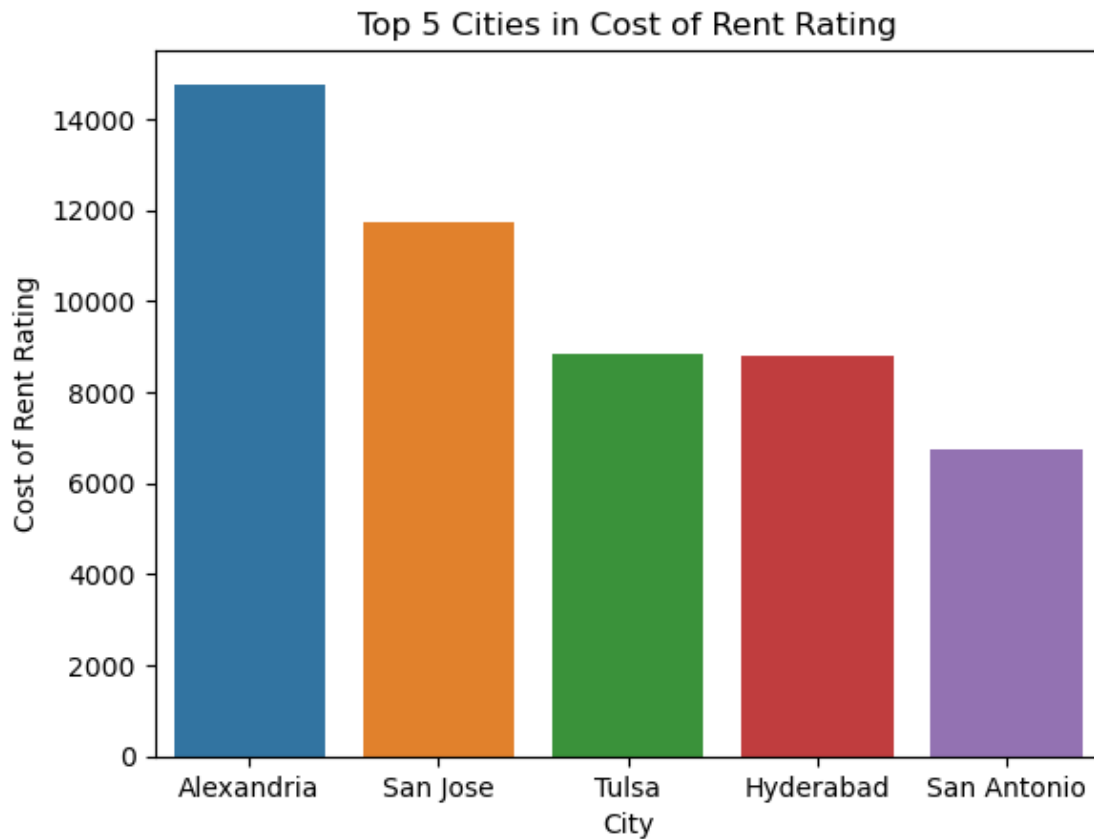
[9]: Text(0.5, 1.0, 'Top 5 Cities in Cost of Living Rating')

```



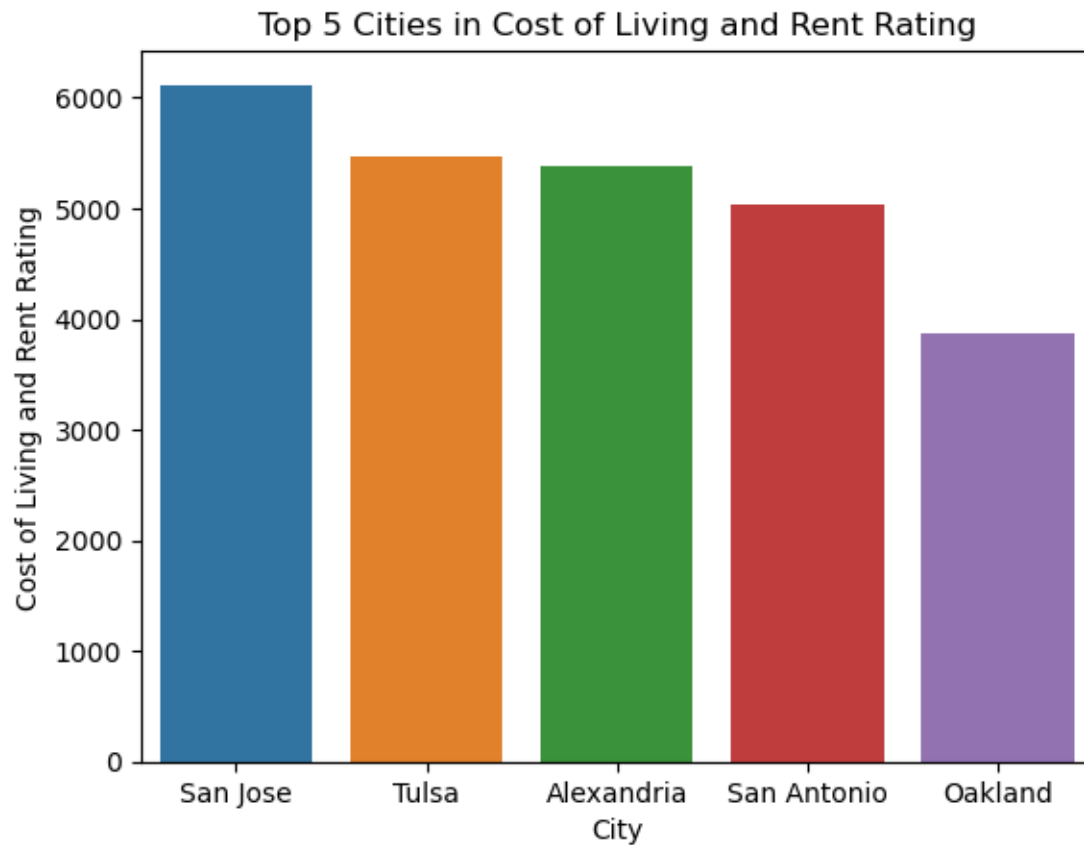
```
[10]: rent = city_rankings_df.sort_values(by=['rent_rating'], ascending=False).  
      ↪reset_index().loc[0:4, ['city', 'rent_rating']]  
sns.barplot(data=rent, x='city', y='rent_rating')  
plt.xlabel("City")  
plt.ylabel("Cost of Rent Rating")  
plt.title("Top 5 Cities in Cost of Rent Rating")
```

```
[10]: Text(0.5, 1.0, 'Top 5 Cities in Cost of Rent Rating')
```



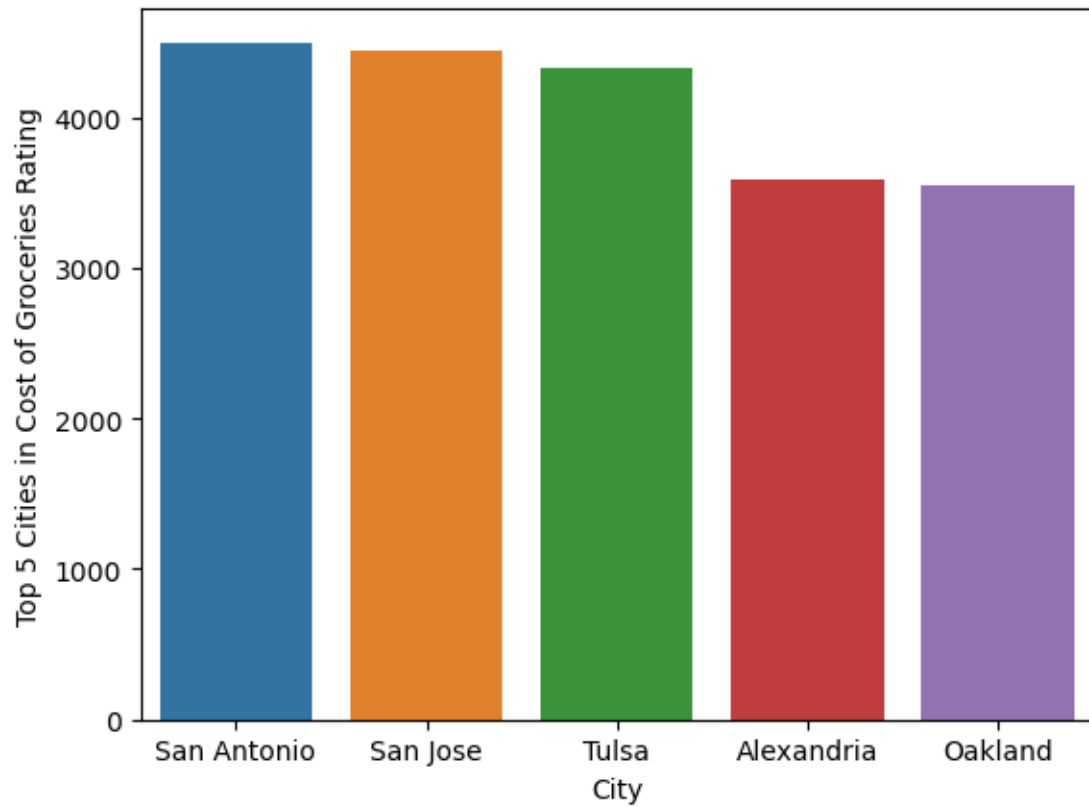
```
[11]: living_rent = city_rankings_df.sort_values(by=['living_rent_rating'],  
        ↪ascending=False).reset_index().loc[0:4, ['city', 'living_rent_rating']]  
sns.barplot(data=living_rent, x='city', y='living_rent_rating')  
plt.xlabel("City")  
plt.ylabel("Cost of Living and Rent Rating")  
plt.title("Top 5 Cities in Cost of Living and Rent Rating")
```

```
[11]: Text(0.5, 1.0, 'Top 5 Cities in Cost of Living and Rent Rating')
```



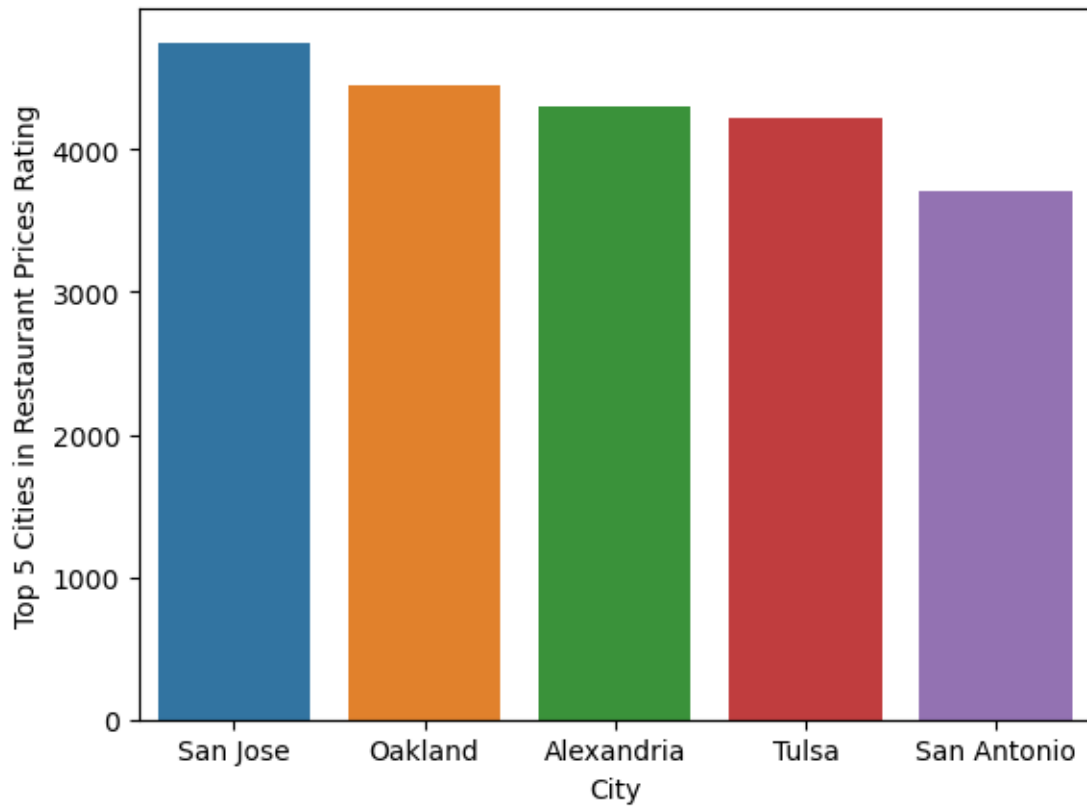
```
[12]: groceries = city_rankings_df.sort_values(by=['groceries_rating'],  
→ascending=False).reset_index().loc[0:4, ['city', 'groceries_rating']]  
sns.barplot(data=groceries, x='city', y='groceries_rating')  
plt.xlabel("City")  
plt.ylabel("Cost of Groceries Rating")  
plt.ylabel("Top 5 Cities in Cost of Groceries Rating")
```

```
[12]: Text(0, 0.5, 'Top 5 Cities in Cost of Groceries Rating')
```



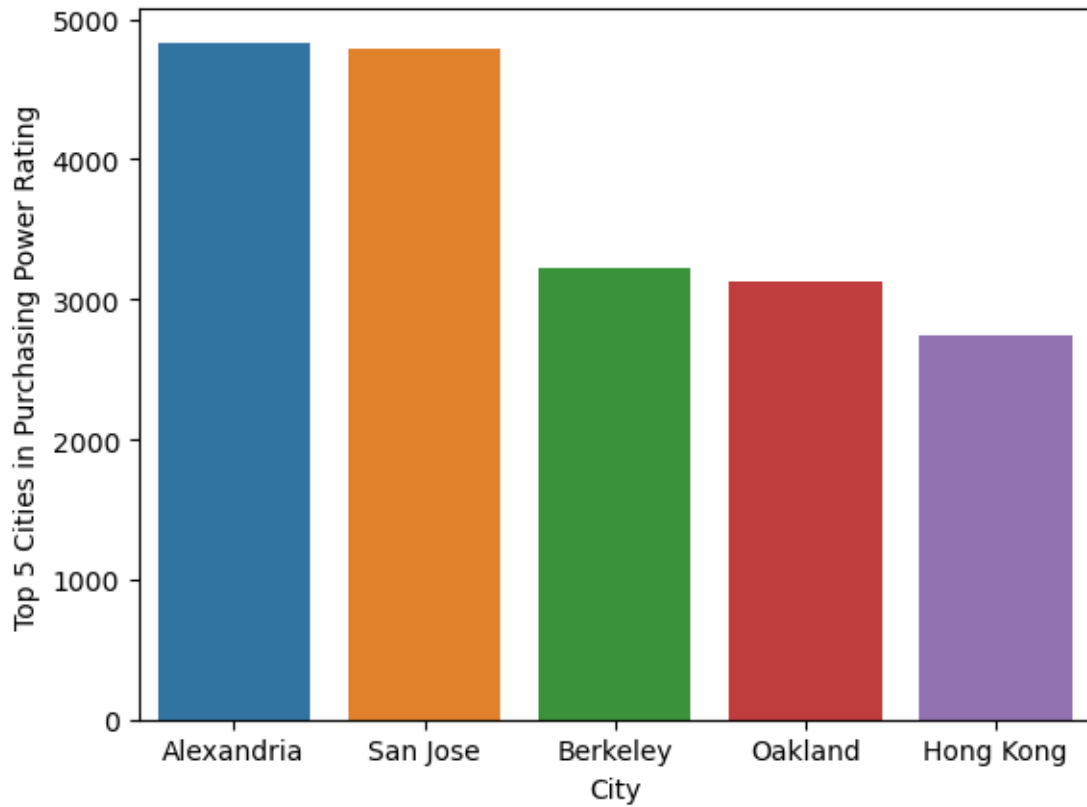
```
[13]: restaurant = city_rankings_df.sort_values(by=['restaurant_rating'],
↪ascending=False).reset_index().loc[0:4, ['city', 'restaurant_rating']]
sns.barplot(data=restaurant, x='city', y='restaurant_rating')
plt.xlabel("City")
plt.ylabel("Restuarant Prices Rating")
plt.ylabel("Top 5 Cities in Restaurant Prices Rating")
```

```
[13]: Text(0, 0.5, 'Top 5 Cities in Restaurant Prices Rating')
```



```
[14]: purchase = city_rankings_df.sort_values(by=['purchase_rating'], ascending=False).  
      ↪reset_index().loc[0:4, ['city', 'purchase_rating']]  
      sns.barplot(data=purchase, x='city', y='purchase_rating')  
      plt.xlabel("City")  
      plt.ylabel("Purchasing Power Rating")  
      plt.ylabel("Top 5 Cities in Purchasing Power Rating")
```

```
[14]: Text(0, 0.5, 'Top 5 Cities in Purchasing Power Rating')
```

1.2.10 Top 5 Cities by Index

```
[15]: print("Top 5 Cities in Cost of Living:")
      print(cost)
      print("-----")
      print("Top 5 Cities in Cost of Rent:")
      print(rent)
      print("-----")
      print("Top 5 Cities in Cost of Rent and Living:")
      print(living_rent)
      print("-----")
      print("Top 5 Cities in Cost of Groceries:")
      print(groceries)
      print("-----")
      print("Top 5 Cities in Restaurant Prices:")
      print(restaurant)
      print("-----")
      print("Top 5 Cities in Purchase Power:")
      print(purchase)
      print("-----")
```

Top 5 Cities in Cost of Living:

	city	cost_rating
0	San Jose	4302.203568
1	San Antonio	4114.075731
2	Tulsa	4089.815557
3	Oakland	3766.275691
4	Alexandria	3447.062962

Top 5 Cities in Cost of Rent:

	city	rent_rating
0	Alexandria	14759.036145
1	San Jose	11734.401832
2	Tulsa	8817.427386
3	Hyderabad	8809.891808
4	San Antonio	6728.507970

Top 5 Cities in Cost of Rent and Living:

	city	living_rent_rating
0	San Jose	6117.576843
1	Tulsa	5462.724936
2	Alexandria	5378.704720
3	San Antonio	5030.279351
4	Oakland	3866.548829

Top 5 Citits in Cost of Groceries:

	city	groceries_rating
0	San Antonio	4497.816594
1	San Jose	4442.999567
2	Tulsa	4330.106979
3	Alexandria	3585.803147
4	Oakland	3554.743043

Top 5 Cities in Restaurant Prices:

	city	restaurant_rating
0	San Jose	4749.768304
1	Oakland	4446.703087
2	Alexandria	4302.019315
3	Tulsa	4216.269841
4	San Antonio	3707.703384

Top 5 Citites in Purchase Power:

	city	purchase_rating
0	Alexandria	4832.347140
1	San Jose	4783.014466
2	Berkeley	3229.190954
3	Oakland	3132.551687
4	Hong Kong	2738.059020

1.2.11 Determine ratings by country

```
[16]: country_median_df = pd.DataFrame(cleaned_salary.
    ↳groupby(by="country")['salary_in_usd'].median())
country_median_df.reset_index(inplace=True)
cost_country_group = pd.DataFrame(cost_of_living_data.groupby(by="country").
    ↳agg('median'))
cost_country_group.reset_index(inplace=True)

country_rankings_df = pd.merge(country_median_df,
                                cost_country_group,
                                how='outer')
country_rankings_df = country_rankings_df.assign(cost_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Cost of Living
    ↳Index'],

                                rent_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Rent Index'],
                                living_rent_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Cost of Living
    ↳Plus Rent Index'],

                                groceries_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Groceries Index'],
                                restaurant_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Restaurant Price
    ↳Index'],

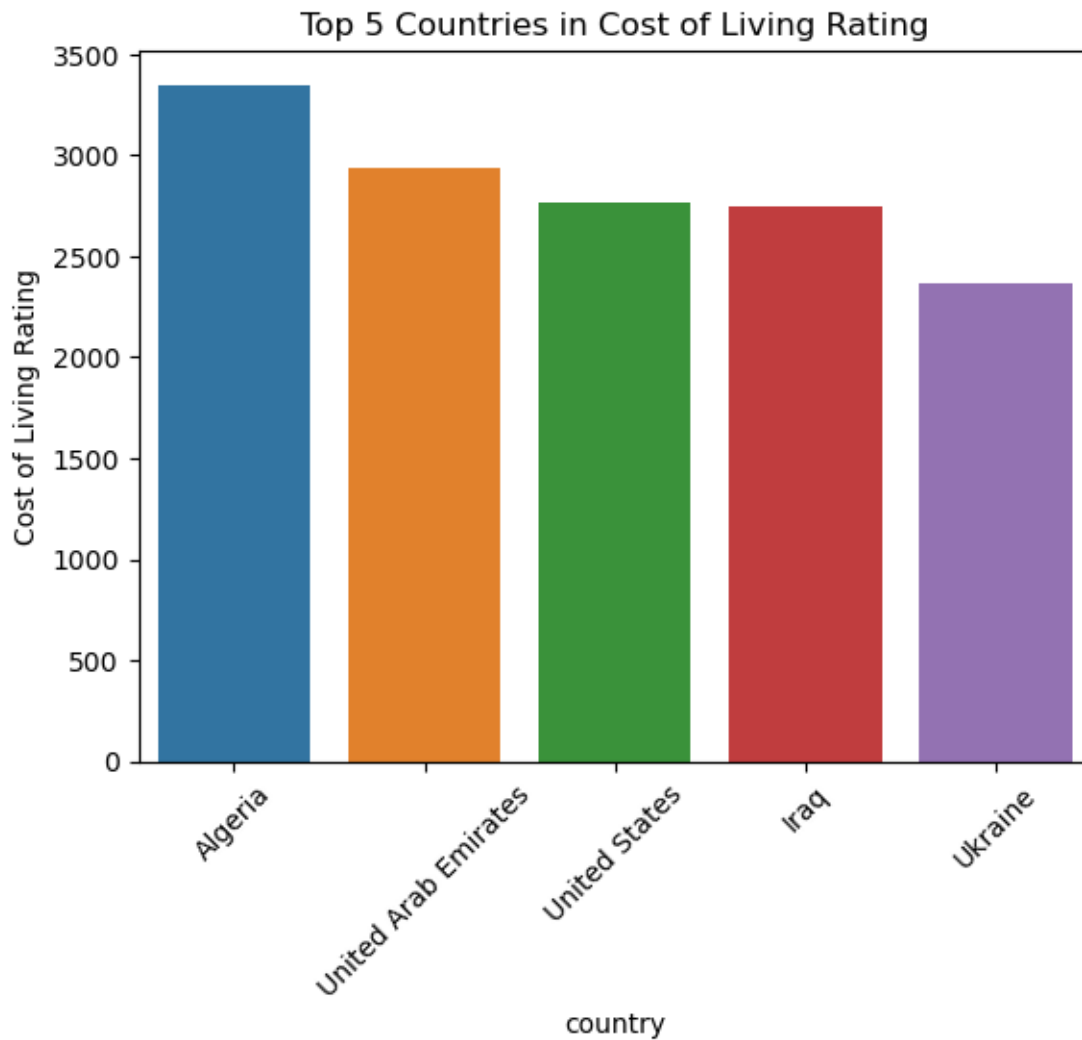
                                purchase_rating = lambda x:
    ↳country_rankings_df['salary_in_usd'] / country_rankings_df['Local Purchasing
    ↳Power Index'])
```

```
C:\Users\janna\AppData\Local\Temp\ipykernel_23848\1076841768.py:3:
FutureWarning: The default value of numeric_only in DataFrameGroupBy.median is
deprecated. In a future version, numeric_only will default to False. Either
specify numeric_only or select only columns which should be valid for the
function.
```

```
cost_country_group =
pd.DataFrame(cost_of_living_data.groupby(by="country").agg('median'))
```

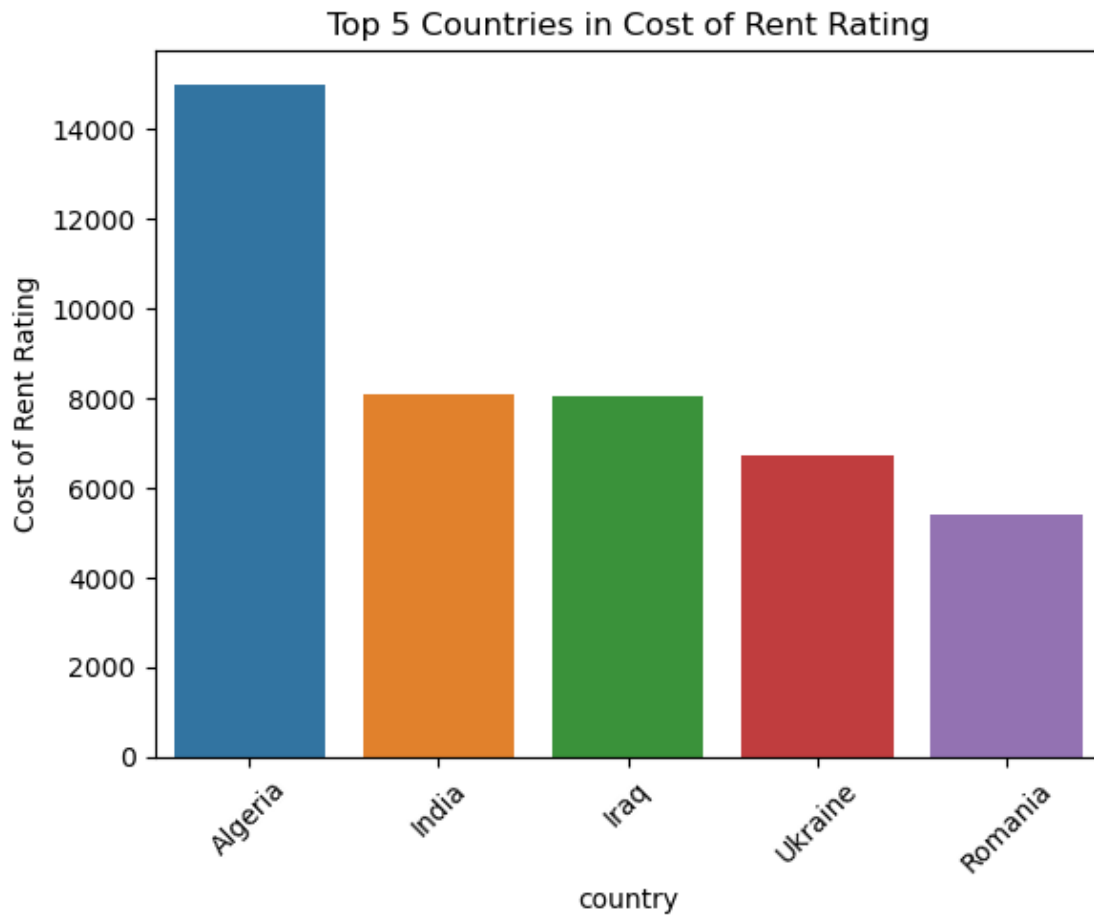
```
[17]: cost = country_rankings_df.sort_values(by=['cost_rating'], ascending=False).
    ↳reset_index().loc[0:4, ['country', 'cost_rating']]
sns.barplot(data=cost, x='country', y='cost_rating')
plt.xlabel("country")
plt.xticks(rotation=45)
plt.ylabel("Cost of Living Rating")
plt.title("Top 5 Countries in Cost of Living Rating")
```

```
[17]: Text(0.5, 1.0, 'Top 5 Countries in Cost of Living Rating')
```



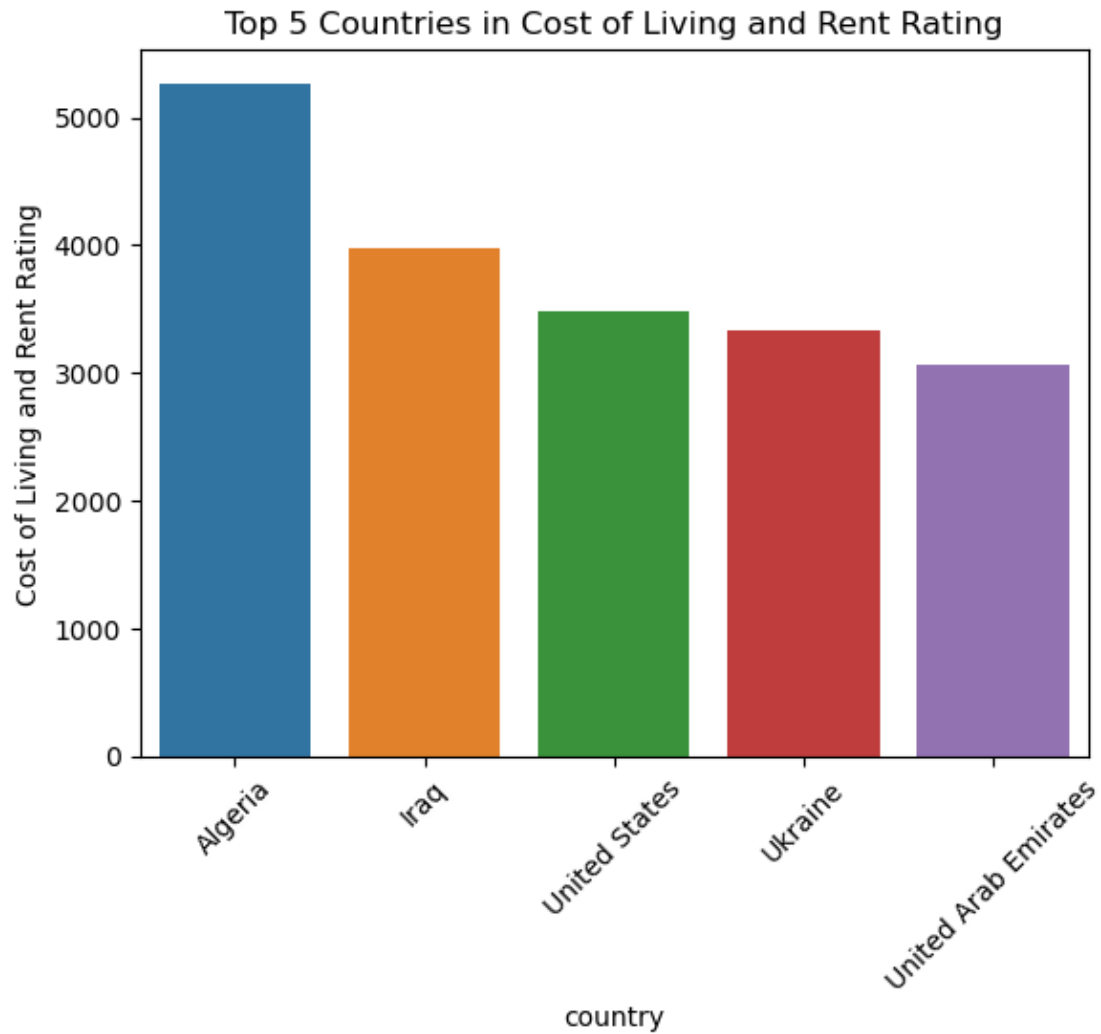
```
[18]: rent = country_rankings_df.sort_values(by=['rent_rating'], ascending=False).  
      ↪reset_index().loc[0:4, ['country', 'rent_rating']]  
sns.barplot(data=rent, x='country', y='rent_rating')  
plt.xlabel("country")  
plt.xticks(rotation=45)  
plt.ylabel("Cost of Rent Rating")  
plt.title("Top 5 Countries in Cost of Rent Rating")
```

```
[18]: Text(0.5, 1.0, 'Top 5 Countries in Cost of Rent Rating')
```



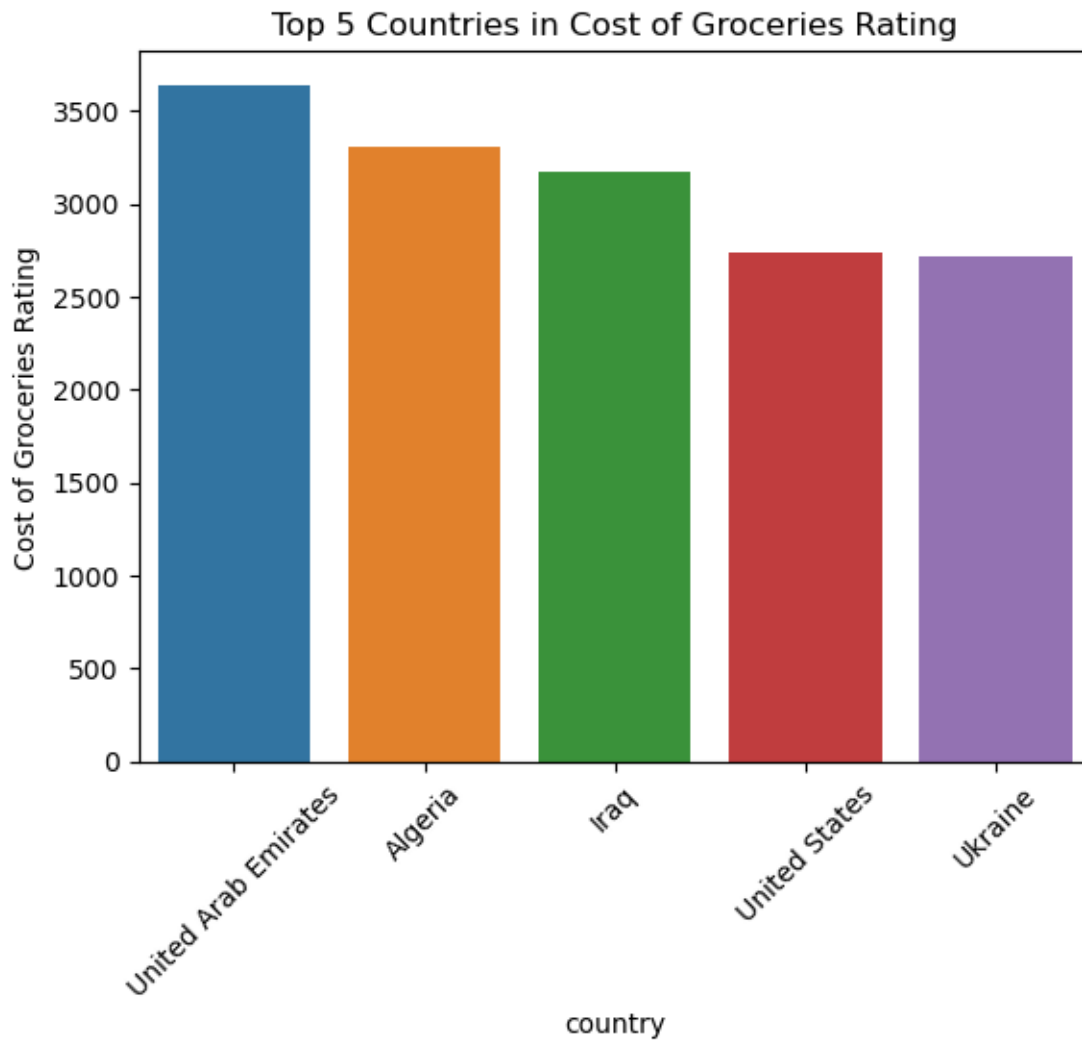
```
[19]: living_rent = country_rankings_df.sort_values(by=['living_rent_rating'],  
↪ascending=False).reset_index().loc[0:4, ['country', 'living_rent_rating']]  
sns.barplot(data=living_rent, x='country', y='living_rent_rating')  
plt.xlabel("country")  
plt.xticks(rotation=45)  
plt.ylabel("Cost of Living and Rent Rating")  
plt.title("Top 5 Countries in Cost of Living and Rent Rating")
```

```
[19]: Text(0.5, 1.0, 'Top 5 Countries in Cost of Living and Rent Rating')
```



```
[20]: groceries = country_rankings_df.sort_values(by=['groceries_rating'],  
        ↪ascending=False).reset_index().loc[0:4, ['country', 'groceries_rating']]  
sns.barplot(data=groceries, x='country', y='groceries_rating')  
plt.xlabel("country")  
plt.xticks(rotation=45)  
plt.ylabel("Cost of Groceries Rating")  
plt.title("Top 5 Countries in Cost of Groceries Rating")
```

```
[20]: Text(0.5, 1.0, 'Top 5 Countries in Cost of Groceries Rating')
```



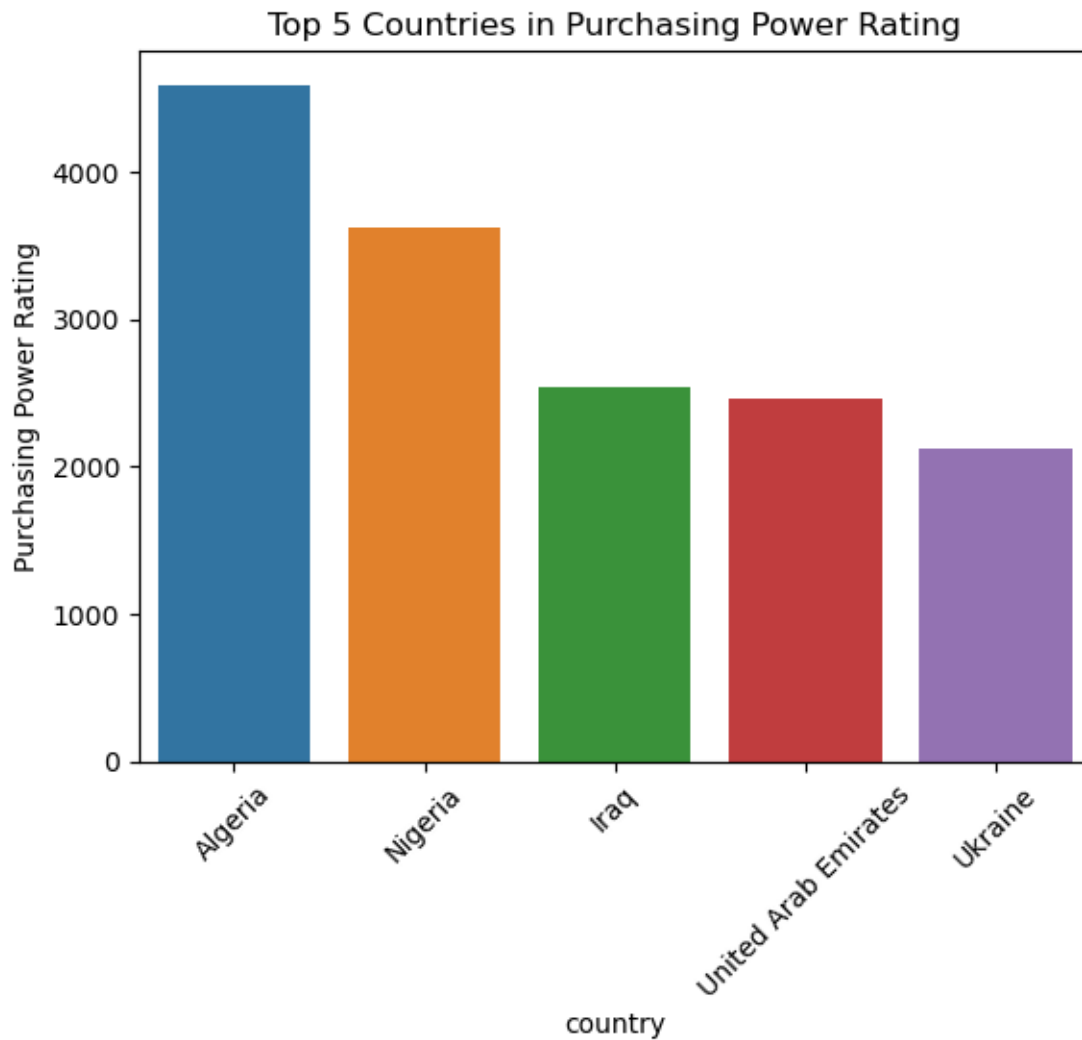
```
[21]: restaurant = country_rankings_df.sort_values(by=['restaurant_rating'],
↪ascending=False).reset_index().loc[0:4, ['country', 'restaurant_rating']]
sns.barplot(data=restaurant, x='country', y='restaurant_rating')
plt.xlabel("country")
plt.xticks(rotation=45)
plt.ylabel("Restaurant Prices Ratings")
plt.title("Top 5 Countries in Restaurant Prices Rating")
```

```
[21]: Text(0.5, 1.0, 'Top 5 Countries in Restaurant Prices Rating')
```



```
[22]: purchase = country_rankings_df.sort_values(by=['purchase_rating'],  
        ↪ascending=False).reset_index().loc[0:4, ['country', 'purchase_rating']]  
sns.barplot(data=purchase, x='country', y='purchase_rating')  
plt.xlabel("country")  
plt.xticks(rotation=45)  
plt.ylabel("Purchasing Power Rating")  
plt.title("Top 5 Countries in Purchasing Power Rating")
```

```
[22]: Text(0.5, 1.0, 'Top 5 Countries in Purchasing Power Rating')
```

1.2.12 Top 5 Countries by Index

```
[23]: print("Top 5 Countries in Cost of Living:")
      print(cost)
      print("-----")
      print("Top 5 Countries in Cost of Rent:")
      print(rent)
      print("-----")
      print("Top 5 Countries in Cost of Rent and Living:")
      print(living_rent)
      print("-----")
      print("Top 5 Countries in Cost of Groceries:")
      print(groceries)
      print("-----")
```

```

print("Top 5 Countries in Restaurant Prices:")
print(restaurant)
print("-----")
print("Top 5 Countries in Purchase Power:")
print(purchase)
print("-----")

```

Top 5 Countries in Cost of Living:

	country	cost_rating
0	Algeria	3351.206434
1	United Arab Emirates	2941.975197
2	United States	2766.908888
3	Iraq	2745.744097
4	Ukraine	2363.755745

Top 5 Countries in Cost of Rent:

	country	rent_rating
0	Algeria	14992.503748
1	India	8088.922156
2	Iraq	8032.128514
3	Ukraine	6703.910615
4	Romania	5390.835580

Top 5 Countries in Cost of Rent and Living:

	country	living_rent_rating
0	Algeria	5268.703899
1	Iraq	3970.617431
2	United States	3481.624758
3	Ukraine	3330.249769
4	United Arab Emirates	3060.551841

Top 5 Countries in Cost of Groceries:

	country	groceries_rating
0	United Arab Emirates	3641.456583
1	Algeria	3305.785124
2	Iraq	3173.595684
3	United States	2734.806630
4	Ukraine	2719.033233

Top 5 Countries in Restaurant Prices:

	country	restaurant_rating
0	Algeria	4810.004810
1	Iraq	3397.893306
2	Ukraine	2925.639984
3	United States	2784.026997
4	United Arab Emirates	2752.837540

Top 5 Countries in Purchase Power:

	country	purchase_rating
0	Algeria	4591.368228
1	Nigeria	3614.457831
2	Iraq	2534.533012
3	United Arab Emirates	2463.427575
4	Ukraine	2118.893467

[]: