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

In [3]: data= pd.read_csv(r'D:/prodigy/task1 world population/Metadata_Country_API_SP.POP.T
In [4]: data
```

Out[4]:		Country Code	Region	IncomeGroup	SpecialNotes	TableName	Unnamed: 5
	0	ABW	Latin America & Caribbean	High income	NaN	Aruba	NaN
	1	AFE	NaN	NaN	26 countries, stretching from the Red Sea in t	Africa Eastern and Southern	NaN
	2	AFG	South Asia	Low income	The reporting period for national accounts dat	Afghanistan	NaN
	3	AFW	NaN	NaN	22 countries, stretching from the westernmost 	Africa Western and Central	NaN
	4	AGO	Sub- Saharan Africa	Lower middle income	The World Bank systematically assesses the app	Angola	NaN
	•••	•••	•••			•••	
	260	XKX	Europe & Central Asia	Upper middle income	NaN	Kosovo	NaN
	261	YEM	Middle East & North Africa	Low income	The World Bank systematically assesses the app	Yemen, Rep.	NaN
	262	ZAF	Sub- Saharan Africa	Upper middle income	Fiscal year end: March 31; reporting period fo	South Africa	NaN
	263	ZMB	Sub- Saharan Africa	Lower middle income	National accounts data were rebased to reflect	Zambia	NaN
	264	ZWE	Sub- Saharan Africa	Lower middle income	National Accounts data are reported in Zimbabw	Zimbabwe	NaN

265 rows × 6 columns

```
In [5]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np

region_counts = data['Region'].value_counts()

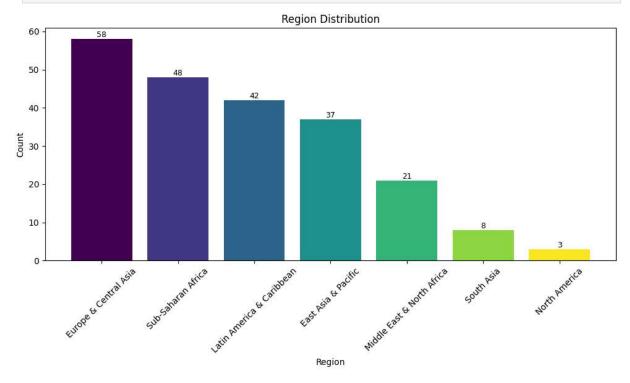
colors = plt.cm.viridis(np.linspace(0, 1, len(region_counts.index)))
```

```
plt.figure(figsize=(10, 6))
bars = plt.bar(region_counts.index, region_counts.values, color=colors)

plt.xlabel('Region')
plt.ylabel('Count')
plt.title('Region Distribution')

plt.xticks(rotation=45)

for bar in bars:
    plt.text(bar.get_x() + bar.get_width()/2, bar.get_height(), round(bar.get_height ha='center', va='bottom', fontsize=9)
plt.tight_layout()
plt.show()
```



Column Non-Null Count Dtype Country Code 265 non-null object 1 217 non-null object Region 2 IncomeGroup 216 non-null object 3 SpecialNotes 127 non-null object 4 TableName 265 non-null object 5 Unnamed: 5 0 non-null float64

dtypes: float64(1), object(5)
memory usage: 12.6+ KB

```
data.describe()
 In [8]:
 Out[8]:
                Unnamed: 5
         count
                        0.0
                       NaN
          mean
            std
                       NaN
           min
                       NaN
           25%
                       NaN
           50%
                       NaN
          75%
                       NaN
           max
                       NaN
 In [9]:
         data.isnull().sum()
 Out[9]: Country Code
                           0
                          48
         Region
         IncomeGroup
                          49
         SpecialNotes
                         138
         TableName
                           0
         Unnamed: 5
                         265
         dtype: int64
In [10]: data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 265 entries, 0 to 264
        Data columns (total 6 columns):
                          Non-Null Count Dtype
             Column
             ----
                           -----
                                           object
         0
             Country Code 265 non-null
         1
             Region
                           217 non-null
                                           object
         2
             IncomeGroup
                           216 non-null
                                           object
         3
             SpecialNotes 127 non-null
                                           object
             TableName
                           265 non-null
                                           object
                                           float64
             Unnamed: 5
                           0 non-null
        dtypes: float64(1), object(5)
        memory usage: 12.6+ KB
In [11]: data['Region'].fillna(data['Region'], inplace=True)
         data['IncomeGroup'].fillna(data['IncomeGroup'], inplace=True)
In [12]: data['SpecialNotes'] = data['SpecialNotes'].fillna('No Special Notes')
In [13]: data = data.rename(columns={'Unnamed: 5': 'Gender'})
         data['Gender'] = data['Gender'].fillna('others')
In [14]:
         data
```

Out[14]:		Country Code	Region	IncomeGroup	SpecialNotes	TableName	Gender
	0	ABW	Latin America & Caribbean	High income	No Special Notes	Aruba	others
	1	AFE	NaN	NaN	26 countries, stretching from the Red Sea in t	Africa Eastern and Southern	others
	2	AFG	South Asia	Low income	The reporting period for national accounts dat	Afghanistan	others
	3	AFW	NaN	NaN	22 countries, stretching from the westernmost	Africa Western and Central	others
	4	AGO	Sub-Saharan Africa	Lower middle income	The World Bank systematically assesses the app	Angola	others
	•••						•••
	260	XKX	Europe & Central Asia	Upper middle income	No Special Notes	Kosovo	others
	261	YEM	Middle East & North Africa	Low income	The World Bank systematically assesses the app	Yemen, Rep.	others
	262	ZAF	Sub-Saharan Africa	Upper middle income	Fiscal year end: March 31; reporting period fo	South Africa	others
	263	ZMB	Sub-Saharan Africa	Lower middle income	National accounts data were rebased to reflect	Zambia	others
	264	ZWE	Sub-Saharan Africa	Lower middle income	National Accounts data are reported in Zimbabw	Zimbabwe	others
	265 ro	ws x 6 col	umns				

265 rows × 6 columns

```
In [15]: data = data.rename(columns={'Unnamed: 5': 'Gender'})
    data
```

Out[15]:		Country Code	Region	IncomeGroup	SpecialNotes	TableName	Gender
	0	ABW	Latin America & Caribbean	High income	No Special Notes	Aruba	others
	1	AFE	NaN	NaN	26 countries, stretching from the Red Sea in t	Africa Eastern and Southern	others
	2	AFG	South Asia	Low income	The reporting period for national accounts dat	Afghanistan	others
	3	AFW	NaN	NaN	22 countries, stretching from the westernmost	Africa Western and Central	others
	4	AGO	Sub-Saharan Africa	Lower middle income	The World Bank systematically assesses the app	Angola	others
	•••						
	260	XKX	Europe & Central Asia	Upper middle income	No Special Notes	Kosovo	others
	261	YEM	Middle East & North Africa	Low income	The World Bank systematically assesses the app	Yemen, Rep.	others
	262	ZAF	Sub-Saharan Africa	Upper middle income	Fiscal year end: March 31; reporting period fo	South Africa	others
	263	ZMB	Sub-Saharan Africa	Lower middle income	National accounts data were rebased to reflect	Zambia	others
	264	ZWE	Sub-Saharan Africa	Lower middle income	National Accounts data are reported in Zimbabw	Zimbabwe	others

265 rows × 6 columns

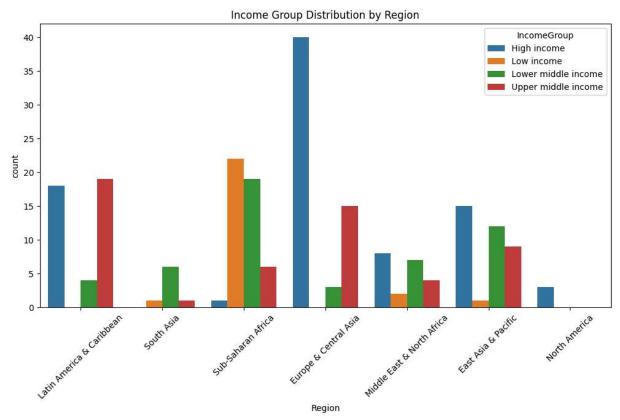
In [16]: data

out[16]:		Country Code	Region	IncomeGroup	SpecialNotes	TableName	Gender
	0	ABW	Latin America & Caribbean	High income	No Special Notes	Aruba	others
	1	AFE	NaN	NaN	26 countries, stretching from the Red Sea in t	Africa Eastern and Southern	others
	2	AFG	South Asia	Low income	The reporting period for national accounts dat	Afghanistan	others
	3	AFW	NaN	NaN	22 countries, stretching from the westernmost	Africa Western and Central	others
	4	AGO	Sub-Saharan Africa	Lower middle income	The World Bank systematically assesses the app	Angola	others
	•••	•••					
	260	XKX	Europe & Central Asia	Upper middle income	No Special Notes	Kosovo	others
	261	YEM	Middle East & North Africa	Low income	The World Bank systematically assesses the app	Yemen, Rep.	others
	262	ZAF	Sub-Saharan Africa	Upper middle income	Fiscal year end: March 31; reporting period fo	South Africa	others
	263	ZMB	Sub-Saharan Africa	Lower middle income	National accounts data were rebased to reflect	Zambia	others
	264	ZWE	Sub-Saharan Africa	Lower middle income	National Accounts data are reported in Zimbabw	Zimbabwe	others
	265 ro	ws × 6 col	umns				

265 rows × 6 columns

```
import seaborn as sns
import matplotlib.pyplot as plt

# Count plot for Income Group by Region
plt.figure(figsize=(12, 6))
sns.countplot(data=data, x='Region', hue='IncomeGroup')
plt.title('Income Group Distribution by Region')
plt.xticks(rotation=45)
plt.show()
```



```
In [19]: from wordcloud import WordCloud

text = ' '.join(data['SpecialNotes'].dropna())

wordcloud = WordCloud(width=800, height=400, background_color='white').generate(tex

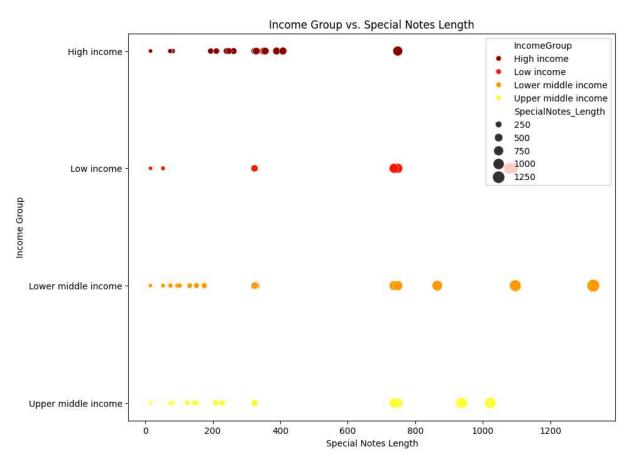
plt.figure(figsize=(10, 5))
 plt.imshow(wordcloud, interpolation='bilinear')
 plt.axis('off')
 plt.title('Word Cloud of Special Notes')
 plt.show()
```

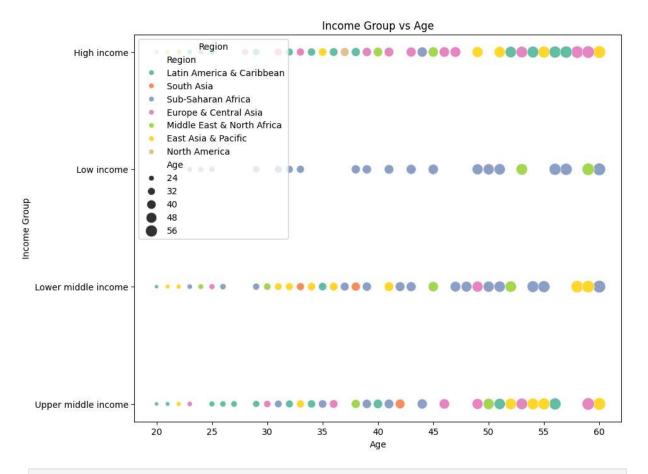
Word Cloud of Special Notes



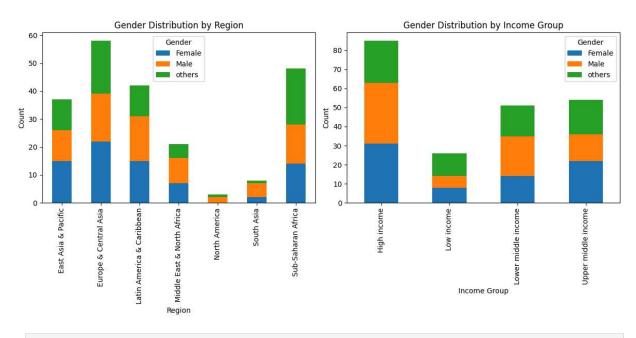
```
In [20]: data['SpecialNotes_Length'] = data['SpecialNotes'].str.len()

# Plotting the scatter plot
plt.figure(figsize=(10, 8))
sns.scatterplot(x='SpecialNotes_Length', y='IncomeGroup', hue='IncomeGroup', data=d
plt.title('Income Group vs. Special Notes Length')
plt.xlabel('Special Notes Length')
plt.ylabel('Income Group')
plt.show()
```





```
In [23]: genders = ['Male', 'Female', 'others']
         data['Gender'] = np.random.choice(genders, size=len(data))
         # Plotting bar chart for Gender vs Region
         plt.figure(figsize=(12, 6))
         plt.subplot(1, 2, 1)
         region_gender_counts = data.groupby(['Region', 'Gender']).size().unstack(fill value
         region_gender_counts.plot(kind='bar', stacked=True, ax=plt.gca())
         plt.title('Gender Distribution by Region')
         plt.xlabel('Region')
         plt.ylabel('Count')
         # Plotting bar chart for Gender vs IncomeGroup
         plt.subplot(1, 2, 2)
         income_gender_counts = data.groupby(['IncomeGroup', 'Gender']).size().unstack(fill_
         income_gender_counts.plot(kind='bar', stacked=True, ax=plt.gca())
         plt.title('Gender Distribution by Income Group')
         plt.xlabel('Income Group')
         plt.ylabel('Count')
         plt.tight_layout()
         plt.show()
```



In []: