25	19020000	19020000	19020000	19020000	410260140
26	12000000	12000000	12000000	11880000	359880000
27	17520000	18000000	18000000	17520000	417242040
28	214260000	218580000	221100000	226860000	4536540000
29	30000	30000	30000	30000	1123140
30	120000	120000	120000	120000	9970800
31	23400000	23700000	23700000	23580000	590880000
32	3000000	3000000	3000000	3000000	86730000
33	540000	540000	540000	540000	9203040
34	22200000	22500000	22500000	21000000	471850680
35	86400000	88200000	88500000	87000000	2093460000
36	540000	540000	540000	540000	18688020
37	3600000	3900000	4200000	4290000	95190000
38	9000000	9300000	9300000	9420000	157980000
39	300000	300000	300000	300000	8640000
40	141600000	144000000	147000000	145500000	3189660000
41	13003680	13080000	13200000	12600000	299700300
42	2400000	2400000	2400000	2400000	70740000
43	4020000	4020000	4020000	4020000	122916960
44	300000	300000	300000	300000	10080000
45	2100000	2100000	2100000	2100000	58300020
46	81000000	82500000	84000000	84000000	1248600000
47	15240	15240	15240	15600	2167620
48	600000	600000	600000	600000	21090000
49	14400000	14688000	15000000	15240000	284816400
50	99000000	96000000	93000000	76500000	2386067999
51	144000000	150000000	156000000	159000000	1920928320
52	6000000	5400000	3900000	3840000	121620000

```
# Dropping rows where any column contains the value "0"
df = df[(df != 0).all(axis=1)]

# Reset the index
df = df.reset_index(drop=True)
```

<sup>#</sup> Check the resulting DataFrame
print(df)

00000	00000	00000	00000	2540000
				3608400
				35100000
				402000000
				2139960
				8595960
				5360040
				384006000
				642823380
				600600000
				2807280000
				76425060
				143450940
				24794400
				2536776384
				665335200
				410260140
				359880000
				417242040
				4536540000
	30000			1123140
120000	120000	120000	120000	9970800
23400000	23700000	23700000	23580000	590880000
3000000	3000000	3000000	3000000	86730000
22200000	22500000	22500000	21000000	471850680
86400000	88200000	88500000	87000000	2093460000
540000	540000	540000	540000	18688020
3600000	3900000	4200000	4290000	95190000
300000	300000	300000	300000	8640000
141600000	144000000	147000000	145500000	3189660000
13003680	13080000	13200000	12600000	299700300
2400000	2400000	2400000	2400000	70740000
4020000	4020000	4020000	4020000	122916960
300000	300000	300000	300000	10080000
2100000	2100000	2100000	2100000	58300020
81000000	82500000	84000000	84000000	1248600000
15240	15240	15240	15600	2167620
600000	600000	600000	600000	21090000
14400000	14688000	15000000	15240000	284816400
99000000	96000000	93000000	76500000	2386067999
144000000	150000000	156000000	159000000	1920928320
	120000 1200000 15000000 60000 240000 180000 23400000 23400000 181800000 1200000 1200000 1200000 1200000 17520000 1200000 23400000 23400000 23400000 23400000 300000 22200000 86400000 300000 13003680 2400000 13003680 2400000 15240 600000 15240 600000 14400000 99000000	120000       120000         1200000       1200000         15000000       15000000         60000       60000         240000       240000         180000       180000         13200000       23400000         23400000       23400000         20400000       20400000         181800000       190800000         5478000       5500020         4620000       4680000         1200000       1200000         104186484       107596260         19767420       21200040         19020000       1200000         17520000       18000000         214260000       218580000         30000       30000         23400000       23700000         300000       300000         23400000       23700000         86400000       38200000         540000       3900000         300000       300000         141600000       14400000         4020000       4020000         4020000       4020000         4020000       300000         2100000       15240         600000       1468800	120000         120000         120000           1200000         1200000         1200000           15000000         15000000         15000000           60000         60000         60000           240000         240000         240000           180000         180000         180000           13200000         13200000         13200000           23400000         23400000         23400000           20400000         20400000         20700000           181800000         190800000         198000000           5478000         5500020         5520000           4620000         4680000         4680000           1200000         1200000         1200000           19767420         21200040         21900000           19020000         19020000         19020000           12000000         12000000         12000000           17520000         18000000         18000000           214260000         218580000         221100000           23400000         23700000         23700000           23400000         23700000         23700000           300000         300000         300000           3640000	120000         120000         120000         120000           1200000         1200000         1200000         1200000           15000000         15000000         15000000         15000000           60000         60000         60000         60000           240000         240000         240000         240000           180000         180000         180000         1296000           13200000         13200000         12960000         2340000         22620000           23400000         23400000         23400000         22620000         20700000         20700000           2440000         20400000         20700000         20700000         20700000         20700000           18180000         190800000         198000000         195000000         195000000         195000000           4620000         4680000         4680000         4680000         4620000         1200000           1200000         1200000         1200000         1200000         121486440         19767420         21200040         21900000         12120000         1200000         1200000         1200000         1200000         1200000         1200000         12380000         300000         300000         30000

df.head()

	Country	Coffee type	1990/91	1991/92	1992/93	1993/94	1994/95
0	Angola	Robusta/Arabica	1200000	1800000	2100000	1200000	1500000
1	Bolivia (Plurinational State of)	Arabica	1500000	1620000	1650000	1710000	1770000
2	Brazil	Arabica/Robusta	492000000	510000000	534000000	546000000	558000000
3	Burundi	Arabica/Robusta	120000	96000	102000	114600	120000
4	Ecuador	Arabica/Robusta	21000000	21000000	21000000	21000000	21000000

## df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48 entries, 0 to 47
Data columns (total 33 columns):

Column	Non-Null Count	Dtype	
	48 non-null	object	
-		object	
		int64	
1991/92	48 non-null	int64	
1992/93	48 non-null	int64	
1993/94	48 non-null	int64	
1994/95	48 non-null	int64	
1995/96	48 non-null	int64	
1996/97	48 non-null	int64	
1997/98	48 non-null	int64	
1998/99	48 non-null	int64	
1999/00	48 non-null	int64	
2000/01	48 non-null	int64	
2001/02	48 non-null	int64	
2002/03	48 non-null	int64	
		int64	
rotal_domestic_consumption	48 non-null	int64	
	Column Country Coffee type 1990/91 1991/92 1992/93 1993/94 1994/95 1995/96 1996/97 1997/98 1998/99 1999/00 2000/01 2001/02	Column         Non-Null Count           Country         48 non-null           1990/91         48 non-null           1991/92         48 non-null           1992/93         48 non-null           1993/94         48 non-null           1994/95         48 non-null           1995/96         48 non-null           1996/97         48 non-null           1998/99         48 non-null           1999/00         48 non-null           2000/01         48 non-null           2001/02         48 non-null           2002/03         48 non-null           2002/03         48 non-null           2004/05         48 non-null           2005/06         48 non-null           2006/07         48 non-null           2007/08         48 non-null           2009/10         48 non-null           2010/11         48 non-null           2011/12         48 non-null           2015/16         48 non-null           2016/17         48 non-null	

dtypes: int64(31), object(2)
memory usage: 12.5+ KB

## df.describe()

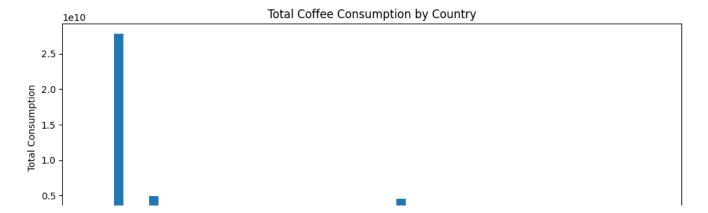
	1990/91	1991/92	1992/93	1993/94	1994/95	1995/96
count	4.800000e+01	4.800000e+01	4.800000e+01	4.800000e+01	4.800000e+01	4.800000e+01
mean	2.438464e+07	2.528325e+07	2.618752e+07	2.637113e+07	2.669833e+07	2.825246e+07
std	7.246522e+07	7.483134e+07	7.835745e+07	7.995455e+07	8.175599e+07	8.861621e+07
min	6.000000e+04	6.000000e+04	4.800000e+04	4.800000e+04	6.000000e+04	4.800000e+04
25%	5.250000e+05	6.825000e+05	8.250000e+05	9.150000e+05	9.000000e+05	7.620000e+05
50%	3.890010e+06	4.260000e+06	3.978000e+06	4.260000e+06	4.700010e+06	4.700010e+06
75%	1.830000e+07	1.830000e+07	1.935000e+07	1.935000e+07	1.890000e+07	2.010000e+07
max	4.920000e+08	5.100000e+08	5.340000e+08	5.460000e+08	5.580000e+08	6.060000e+08

There's steady increase in average consumption of coffee annually.

The significant standard deviations indicate variability in consumption levels from year to year. This variability could be influenced by factors such as economic conditions, coffee type, price fluctuations, or changes in consumer behavior.

```
import matplotlib.pyplot as plt

# Plotting
plt.figure(figsize=(10, 6))
plt.bar(df['Country'], df['Total_domestic_consumption'])
plt.xlabel('Country')
plt.ylabel('Total Consumption')
plt.title('Total Coffee Consumption by Country')
plt.xticks(rotation=90)  # Rotate x-axis labels for better readability
plt.tight_layout()  # Adjust layout to prevent clipping of labels
plt.show()
```



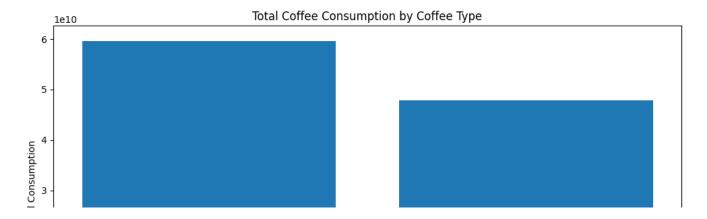
Brazil consumes the most amount of coffee by a very wide margin. Brazil's rich coffee heritage, abundant production, economic significance, diverse culture, and effective promotion contribute to its high levels of coffee consumption.

```
import matplotlib.pyplot as plt

# Split rows with multiple coffee types into separate rows
split_df = df.assign(Coffee_type=df['Coffee type'].str.split('/')).explode('Coffee_type')

# Group by coffee type and calculate total consumption
grouped_df = split_df.groupby('Coffee_type')['Total_domestic_consumption'].sum().reset_ind

# Plotting
plt.figure(figsize=(10, 6))
plt.bar(grouped_df['Coffee_type'], grouped_df['Total_domestic_consumption'])
plt.xlabel('Coffee Type')
plt.ylabel('Total Consumption')
plt.title('Total Coffee Consumption by Coffee Type')
plt.xticks(rotation=45)  # Rotate x-axis labels for better readability
plt.tight_layout()  # Adjust layout to prevent clipping of labels
plt.show()
```



Arabica is the most consumed coffee type

## FURTHER FDA

```
import matplotlib.pyplot as plt

# Extracting the years and minimum consumption
years = df.columns[2:-1]  # Extracting the years from the column names
min_consumption = df.iloc[:, 2:-1].min()  # Calculating the minimum consumption for each y

# Plotting
plt.figure(figsize=(10, 6))
plt.plot(years, min_consumption, marker='o', linestyle='-', color='blue')
plt.xlabel('Years')
plt.ylabel('Minimum Consumption')
plt.title('Minimum Consumption Over the Years')
plt.xticks(rotation=45)  # Rotate x-axis labels for better readability
plt.grid(True)
plt.tight_layout()
plt.show()
```



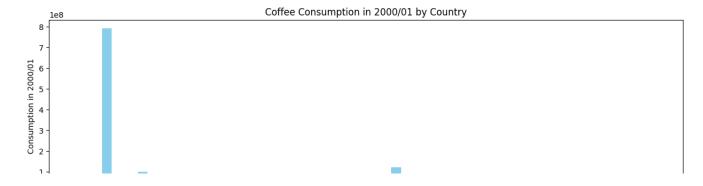
There was a sharp drop in minimum amount of coffee consumed in 2000/01, a sharp rise in 2010/11 and a sharp drop again in 2011/12.

# Extracting the country names and consumption in 2000/01

countries = df['Country']

consumption\_2000\_01 = df['2000/01']

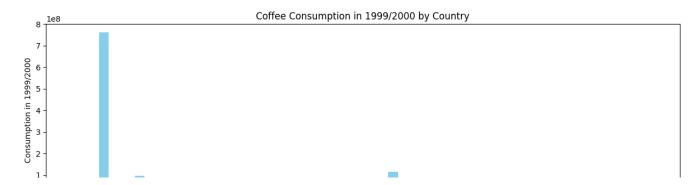
```
# Plotting
plt.figure(figsize=(12, 6))
plt.bar(countries, consumption_2000_01, color='skyblue')
plt.xlabel('Country')
plt.ylabel('Consumption in 2000/01')
plt.title('Coffee Consumption in 2000/01 by Country')
plt.xticks(rotation=90) # Rotate x-axis labels for better readability
plt.tight_layout()
plt.show()
```



```
import matplotlib.pyplot as plt
```

```
# Extracting the country names and consumption in 1999/2000
countries = df['Country']
consumption_1999_2000 = df['1999/00']

# Plotting
plt.figure(figsize=(12, 6))
plt.bar(countries, consumption_1999_2000, color='skyblue')
plt.xlabel('Country')
plt.ylabel('Consumption in 1999/2000')
plt.title('Coffee Consumption in 1999/2000 by Country')
plt.xticks(rotation=90) # Rotate x-axis labels for better readability
plt.tight_layout()
plt.show()
```



Country with the least coffee consumption in 2000/01 was Gabon, this may have been as a result of the fall in coffee price between 2000 and 2004. lower coffee prices could have negatively impacted coffee-producing regions, including Gabon. If coffee farmers in Gabon experienced reduced income due to lower prices, it could have had ripple effects on the local economy, leading to decreased purchasing power and reduced coffee consumption among the population.

I also compared the coffee consumption in all countries for 2000/01 and the past year (1999/00) to check for any change in pattern for other countries. There was no change in pattern, only Gabon was deeply affected.

```
# Selecting columns from '2000/01' to '2011/12'
consumption columns = df.columns[2:-1]
# Dictionary to store the country with the lowest consumption for each year
min_consumption_countries = {}
# Iterating over each year and finding the country with the lowest consumption
for year in consumption_columns:
   min_country = df.loc[df[year].idxmin()]['Country']
   min_consumption_countries[year] = min_country
# Printing the results
for year, country in min_consumption_countries.items():
   print(f"Year: {year}, Country with the lowest consumption: {country}")
     Year: 1990/91, Country with the lowest consumption: Rwanda
     Year: 1991/92, Country with the lowest consumption: Rwanda
     Year: 1992/93, Country with the lowest consumption: Gabon
     Year: 1993/94, Country with the lowest consumption: Gabon
     Year: 1994/95, Country with the lowest consumption: Rwanda
     Year: 1995/96, Country with the lowest consumption: Gabon
     Year: 1996/97, Country with the lowest consumption: Gabon
     Year: 1997/98, Country with the lowest consumption: Gabon
     Year: 1998/99, Country with the lowest consumption: Gabon
     Year: 1999/00, Country with the lowest consumption: Gabon
     Year: 2000/01, Country with the lowest consumption: Gabon
     Year: 2001/02, Country with the lowest consumption: Gabon
     Year: 2002/03, Country with the lowest consumption: Gabon
     Year: 2003/04, Country with the lowest consumption: Gabon
     Year: 2004/05, Country with the lowest consumption: Gabon
     Year: 2005/06, Country with the lowest consumption: Gabon
     Year: 2006/07, Country with the lowest consumption: Gabon
     Year: 2007/08, Country with the lowest consumption: Gabon
     Year: 2008/09, Country with the lowest consumption: Gabon
     Year: 2009/10, Country with the lowest consumption: Gabon
     Year: 2010/11, Country with the lowest consumption: Gabon
     Year: 2011/12, Country with the lowest consumption: Togo
     Year: 2012/13, Country with the lowest consumption: Togo
     Year: 2013/14, Country with the lowest consumption: Togo
     Year: 2014/15, Country with the lowest consumption: Togo
     Year: 2015/16, Country with the lowest consumption: Togo
     Year: 2016/17, Country with the lowest consumption: Togo
     Year: 2017/18, Country with the lowest consumption: Togo
     Year: 2018/19, Country with the lowest consumption: Togo
     Year: 2019/20, Country with the lowest consumption: Togo
```

```
# Filter the DataFrame to get Gabon's row
gabon_row = df[df['Country'] == 'Gabon']
```

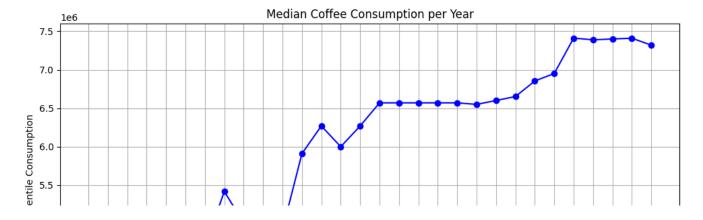
```
# Get Gabon's consumption in 2011/12
gabon_consumption_2011_12 = gabon_row['2011/12'].values[0]
print("Gabon's coffee consumption in 2011/12:", gabon_consumption_2011_12)
    Gabon's coffee consumption in 2011/12: 61920
```

The minimum coffee consumption spiked in 2010/11 with Gabon, then dropped sharply again in 2011/12 with Togo. Togo took the baton as the country with the lowest coffee consumption from 2011/12 till the end date of the dataset (2019/20). Gabon's coffee consumption in 2011/12 was far from the minimum.

```
import matplotlib.pyplot as plt

# Extracting the years and the 50th percentile values
years = df.columns[2:-1]  # Extracting the years from the column names
median_consumption = df.iloc[:, 2:-1].apply(lambda x: x.median())  # Calculating the media

# Plotting
plt.figure(figsize=(10, 6))
plt.plot(years, median_consumption, marker='o', linestyle='-', color='blue')
plt.xlabel('Years')
plt.ylabel('S0th Percentile Consumption')
plt.title('Median Coffee Consumption per Year')
plt.xticks(rotation=45)  # Rotate x-axis labels for better readability
plt.grid(True)
plt.tight_layout()
plt.show()
```



there was a notable drop in coffee consumption in 1999/00 which didn't fully recover until 2001/02. The 50th percentile value (median) dropped significantly. This indicates that half of the countries experienced a decrease in consumption during that period. This drop in consumption could have been influenced by various factors such as the drop in worldwide coffee price in 2000 and the early 2000 economic recession. Surprisingly enough, the 2008 economic recession did not have much effect on median coffee consumption, it had a little bit of effect on the minimum coffee consumption.

```
data = {
    'Country': ['Angola', 'Bolivia', 'Brazil', 'Burundi', 'Ecuador', 'Indonesia', 'Madagas
    'Latitude': [-11.2027, -16.2902, -14.2350, -3.3731, -1.8312, -0.7893, -18.7669, -13.25
    'Longitude': [17.8739, -63.5887, -51.9253, 29.9189, -78.1834, 113.9213, 46.8691, 34.30
}

df2 = pd.DataFrame(data)
  import folium

# Create a map centered around the mean of latitudes and longitudes
  m = folium.Map(location=[df2['Latitude'].mean(), df2['Longitude'].mean()], zoom_start=2)

# Add markers for each country
for index, row in df2.iterrows():
    folium.Marker(location=[row['Latitude'], row['Longitude']], popup=row['Country']).add_

# Display the map
  m
```



Map showing the countries in the dataset

```
import matplotlib.pyplot as plt

# Extracting the years and total consumption
years = df.columns[2:-1]  # Extracting the years from the column names
total_consumption_per_year = df.iloc[:, 2:-1].sum()  # Calculating the total consumption fr

# Plotting
plt.figure(figsize=(10, 6))
plt.plot(years, total_consumption_per_year, marker='o', linestyle='-', color='blue')
plt.xlabel('Years')
plt.ylabel('Total Consumption')
plt.title('Total Consumption per Year')
plt.xticks(rotation=45)  # Rotate x-axis labels for better readability
plt.grid(True)
plt.tight_layout()
plt.show()
```

