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

import seaborn as sns

 $\tt df = pd.read\_csv(r'/content/drive/MyDrive/Prodigy-InfoTech/Task-1/API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_84031 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_84031 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP.POP.TOTL\_DS2\_en\_csv\_v2\_8404 - API\_SP$ 

df

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	196
0	Aruba	ABW	Population, total	SP.POP.TOTL	54608.0	55811.0	56682.
1	Africa Eastern and Southern	AFE	Population, total	SP.POP.TOTL	130692579.0	134169237.0	137835590.
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8622466.0	8790140.0	8969047.
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	97256290.0	99314028.0	101445032.
4	Angola	AGO	Population, total	SP.POP.TOTL	5357195.0	5441333.0	5521400.
							-
261	Kosovo	XKX	Population, total	SP.POP.TOTL	947000.0	966000.0	994000.
262	Yemen, Rep.	YEM	Population, total	SP.POP.TOTL	5542459.0	5646668.0	5753386.
263	South Africa	ZAF	Population, total	SP.POP.TOTL	16520441.0	16989464.0	17503133.
264	Zambia	ZMB	Population, total	SP.POP.TOTL	3119430.0	3219451.0	3323427.
265	Zimbabwe	ZWE	Population, total	SP.POP.TOTL	3806310.0	3925952.0	4049778.
266 rows × 68 columns							

# df.head()

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1962
0	Aruba	ABW	Population, total	SP.POP.TOTL	54608.0	55811.0	56682.0
1	Africa Eastern and Southern	AFE	Population, total	SP.POP.TOTL	130692579.0	134169237.0	137835590.0
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8622466.0	8790140.0	8969047.0
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	97256290.0	99314028.0	101445032.0
4	Angola	AGO	Population, total	SP.POP.TOTL	5357195.0	5441333.0	5521400.0
5 rc	ws × 68 colum	nns					

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1962	
261	Kosovo	XKX	Population, total	SP.POP.TOTL	947000.0	966000.0	994000.0	1
262	Yemen, Rep.	YEM	Population, total	SP.POP.TOTL	5542459.0	5646668.0	5753386.0	Ę
263	South Africa	ZAF	Population, total	SP.POP.TOTL	16520441.0	16989464.0	17503133.0	18
264	Zambia	ZMB	Population, total	SP.POP.TOTL	3119430.0	3219451.0	3323427.0	3
265	Zimbabwe	ZWE	Population, total	SP.POP.TOTL	3806310.0	3925952.0	4049778.0	۷
5 rows × 68 columns								
				SP.POP.TOTL	3806310.0	3925952.0	4049778.0	)

df.shape

(266, 68)

### ${\tt df.columns}$

### df.dtypes

Country Name	object
Country Code	object
Indicator Name	object
Indicator Code	object
1960	float64
2019	float64
2020	float64
2021	float64
2022	float64
2023	float64
Length: 68, dtype	: object

# df.info()

12	1968	264	non-null	float64
13	1969	264	non-null	float64
14	1970	264	non-null	float64
15	1971	264	non-null	float64
16	1972	264	non-null	float64
17	1973	264	non-null	float64
18	1974	264	non-null	float64
19	1975	264	non-null	float64
20	1976	264	non-null	float64
21	1977	264	non-null	float64
22	1978	264	non-null	float64
23	1979	264	non-null	float64
24	1980	264	non-null	float64
25	1981	264	non-null	float64

43	1999	265	non-nu11	†10at64
44	2000	265	non-null	float64
45	2001	265	non-null	float64
46	2002	265	non-null	float64
47	2003	265	non-null	float64
48	2004	265	non-null	float64
49	2005	265	non-null	float64
50	2006	265	non-null	float64
51	2007	265	non-null	float64
52	2008	265	non-null	float64
53	2009	265	non-null	float64
54	2010	265	non-null	float64
55	2011	265	non-null	float64
56	2012	265	non-null	float64
57	2013	265	non-null	float64
58	2014	265	non-null	float64
59	2015	265	non-null	float64
60	2016	265	non-null	float64
61	2017	265	non-null	float64
62	2018	265	non-null	float64
63	2019	265	non-null	float64
64	2020	265	non-null	float64
65	2021	265	non-null	float64
66	2022	265	non-null	float64
67	2023	0 n	on-null	float64
dtyp	es: float64(64),	obje	ect(4)	

dtypes: float64(64), obj memory usage: 141.4+ KB

# df.describe()

	1960	1961	1962	1963	1964	19	
count	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+	
mean	1.172860e+08	1.188956e+08	1.210661e+08	1.237484e+08	1.264530e+08	1.291965e+	
std	3.695500e+08	3.740958e+08	3.808121e+08	3.895098e+08	3.982497e+08	4.071209e+	
min	2.646000e+03	2.888000e+03	3.171000e+03	3.481000e+03	3.811000e+03	4.161000e+	
25%	5.132212e+05	5.231345e+05	5.337595e+05	5.449288e+05	5.566630e+05	5.651150e+	
50%	3.757486e+06	3.887144e+06	4.023896e+06	4.139356e+06	4.224612e+06	4.277636e+	
75%	2.670606e+07	2.748694e+07	2.830289e+07	2.914708e+07	3.001684e+07	3.084892e+	
max	3.031474e+09	3.072422e+09	3.126850e+09	3.193429e+09	3.260442e+09	3.328209e+	
8 rows × 64 columns							

df.duplicated().sum()

0

df.isna().sum().any()

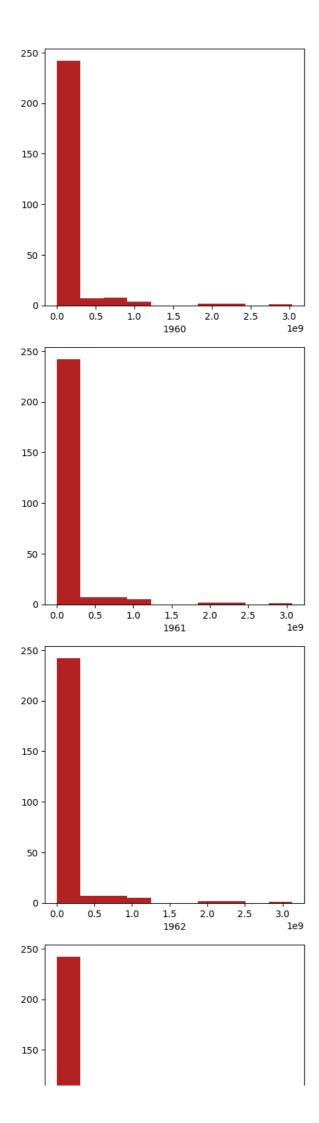
True

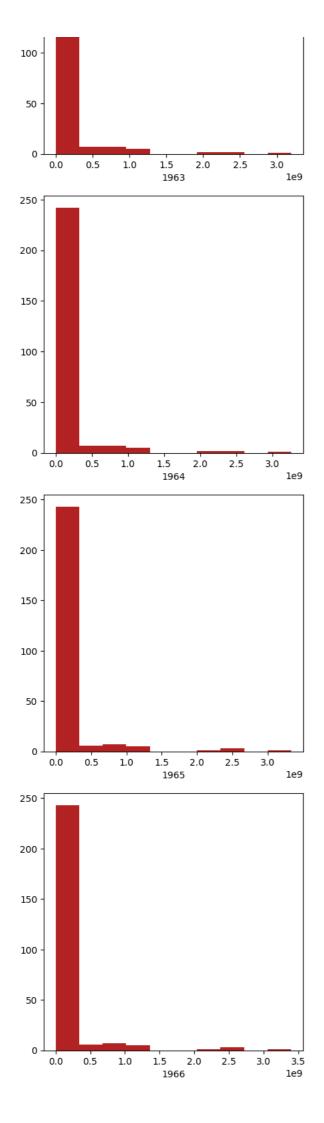
# df=df.fillna(method='ffill') df.head()

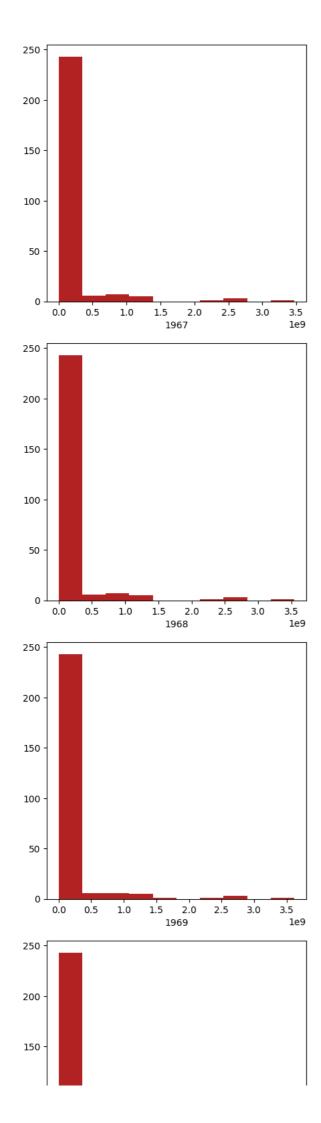
	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1962
0	Aruba	ABW	Population, total	SP.POP.TOTL	54608.0	55811.0	56682.0
1	Africa Eastern and Southern	AFE	Population, total	SP.POP.TOTL	130692579.0	134169237.0	137835590.0
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8622466.0	8790140.0	8969047.0
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	97256290.0	99314028.0	101445032.0
4	Angola	AGO	Population, total	SP.POP.TOTL	5357195.0	5441333.0	5521400.0
5 rc	5 rows × 68 columns						

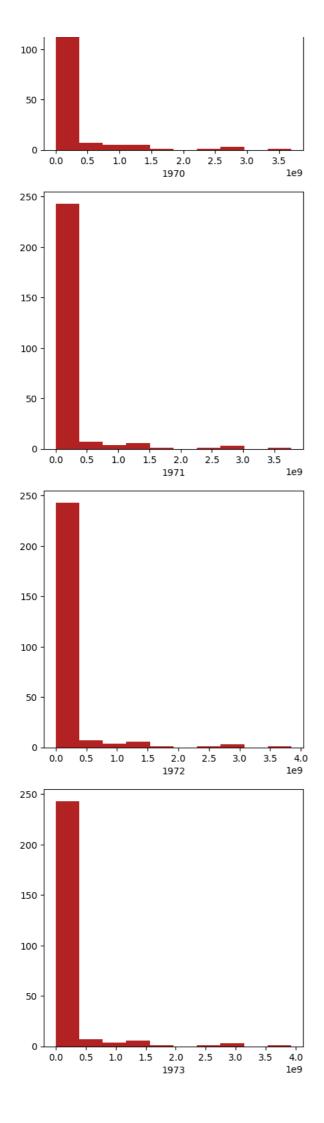
```
df['Country Name'].unique()
                      'Europe & Central Asia (excluding high income)',
                      'Europe & Central Asia', 'Ecuador', 'Egypt, Arab Rep.', 'Euro area', 'Eritrea', 'Spain', 'Estonia', 'Ethiopia',
                     'European Union', 'Fragile and conflict affected situations', 'Finland', 'Fiji', 'France', 'Faroe Islands', 'Micronesia, Fed. Sts.', 'Gabon', 'United Kingdom', 'Georgia', 'Ghana', 'Gibraltar', 'Guinea', 'Gambia, The', 'Guinea-Bissau', 'Equatorial Guinea', 'Greece', 'Grenada', 'Greenland', 'Guatemala', 'Guam', 'Guyana', 'High income', 'Hong Kong SAR, China',
                      'Honduras', 'Heavily indebted poor countries (HIPC)', 'Croatia', 'Haiti', 'Hungary', 'IBRD only', 'IDA & IBRD total', 'IDA total',
                      'IDA blend', 'Indonesia', 'IDA only', 'Isle of Man', 'India',
                     'Not classified', 'Ireland', 'Iran, Islamic Rep.', 'Iraq', 
'Iceland', 'Israel', 'Italy', 'Jamaica', 'Jordan', 'Japan', 
'Kazakhstan', 'Kenya', 'Kyrgyz Republic', 'Cambodia', 'Kiribati', 
'St. Kitts and Nevis', 'Korea, Rep.', 'Kuwait',
                      'Latin America & Caribbean (excluding high income)', 'Lao PDR',
                      'Lebanon', 'Liberia', 'Libya', 'St. Lucia',
                      'Latin America & Caribbean',
                      'Least developed countries: UN classification', 'Low income',
                      'Liechtenstein', 'Sri Lanka', 'Lower middle income',
                      'Low & middle income', 'Lesotho', 'Late-demographic dividend',
                     'Lithuania', 'Luxembourg', 'Latvia', 'Macao SAR, China',
'St. Martin (French part)', 'Morocco', 'Monaco', 'Moldova',
'Madagascar', 'Maldives', 'Middle East & North Africa', 'Mexico',
'Marshall Islands', 'Middle income', 'North Macedonia', 'Mali',
                      'Malta', 'Myanmar'
                     'Maita', 'Myanmar',
'Middle East & North Africa (excluding high income)', 'Montenegro',
'Mongolia', 'Northern Mariana Islands', 'Mozambique', 'Mauritania',
'Mauritius', 'Malawi', 'Malaysia', 'North America', 'Namibia',
'New Caledonia', 'Niger', 'Nigeria', 'Nicaragua', 'Netherlands',
'Norway', 'Nepal', 'Nauru', 'New Zealand', 'OECD members', 'Oman',
'Other small states', 'Pakistan', 'Panama', 'Peru', 'Philippines',
                      'Palau', 'Papua New Guinea', 'Poland', 'Pre-demographic dividend',
                      'Puerto Rico', "Korea, Dem. People's Rep.", 'Portugal', 'Paraguay',
                      'West Bank and Gaza', 'Pacific island small states',
                      'Post-demographic dividend', 'French Polynesia', 'Qatar'
                      'Romania', 'Russian Federation', 'Rwanda', 'South Asia',
'Saudi Arabia', 'Sudan', 'Senegal', 'Singapore', 'Solomon Islands',
'Sierra Leone', 'El Salvador', 'San Marino', 'Somalia', 'Serbia',
'Sub-Saharan Africa (excluding high income)', 'South Sudan',
                      'Sub-Saharan Africa', 'Small states', 'Sao Tome and Principe'
                      'Suriname', 'Slovak Republic', 'Slovenia', 'Sweden', 'Eswatini', 'Sint Maarten (Dutch part)', 'Seychelles', 'Syrian Arab Republic', 'Turks and Caicos Islands', 'Chad',
                      'East Asia & Pacific (IDA & IBRD countries)'
                      'Europe & Central Asia (IDA & IBRD countries)', 'Togo', 'Thailand',
                      'Tajikistan', 'Turkmenistan',
                      'Latin America & the Caribbean (IDA & IBRD countries)',
                      'Timor-Leste', 'Middle East & North Africa (IDA & IBRD countries)',
                      'Tonga', 'South Asia (IDA & IBRD)',
                     'Sub-Saharan Africa (IDA & IBRD countries)', 'Trinidad and Tobago', 'Tunisia', 'Turkiye', 'Tuvalu', 'Tanzania', 'Uganda', 'Ukraine', 'Upper middle income', 'Uruguay', 'United States', 'Uzbekistan', 'St. Vincent and the Grenadines', 'Venezuela, RB',
                      'British Virgin Islands', 'Virgin Islands (U.S.)', 'Viet Nam',
                      'Vanuatu', 'World', 'Samoa', 'Kosovo', 'Yemen, Rep.', 'South Africa', 'Zambia', 'Zimbabwe'], dtype=object)
df['Indicator Name'].unique()
         array(['Population, total'], dtype=object)
df['Indicator Code'].unique()
         array(['SP.POP.TOTL'], dtype=object)
df.drop(['Indicator Name','Indicator Code','Country Code'],axis = 1,inplace = True)
df.columns
         '1976', '1977', '1978', '1979', '1980', '1981', '1982', '1983', '1985', '1986', '1987', '1988', '1989', '1990', '1991', '1992',
                                                                                                                                         '1984'
                     '1994', '1995', '1996', '1997', '1998', '1999', '2000', '2001', '2002', '2003', '2004', '2005', '2006', '2007', '2008', '2008', '2010', '2011', '2012', '2013', '2014', '2015', '2016', '2017', '2018', '2019', '2020',
                      '2021', '2022', '2023'],
```

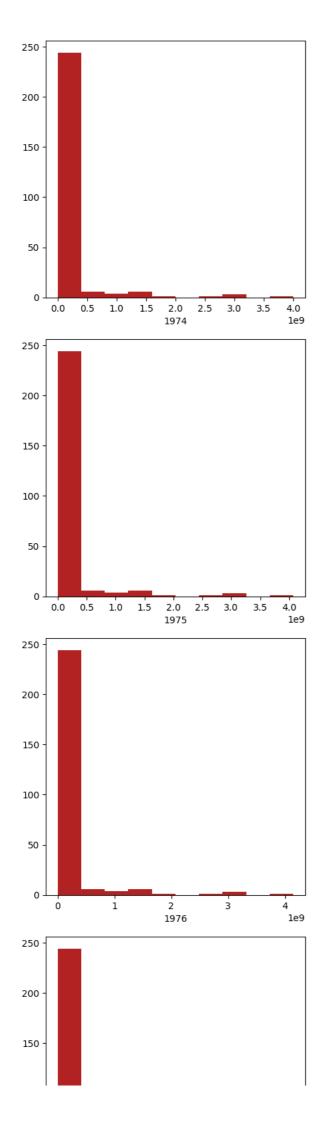
dtype='object')

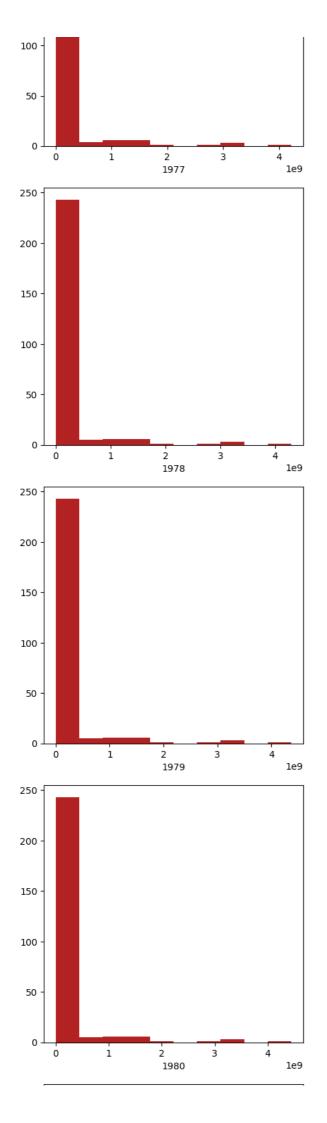


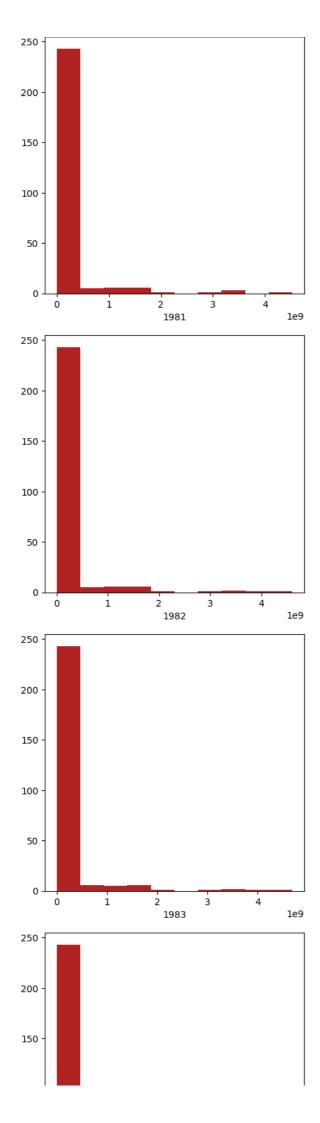


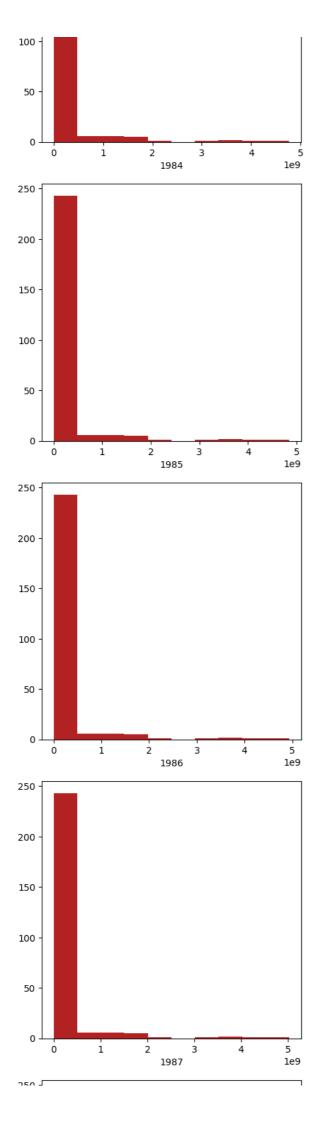


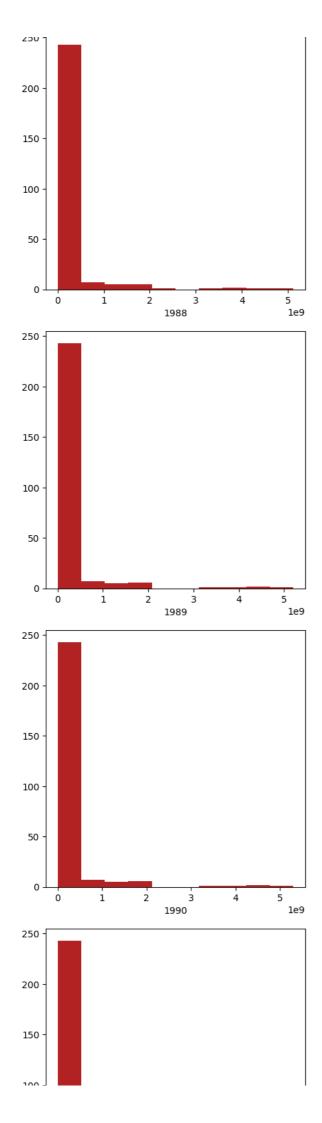


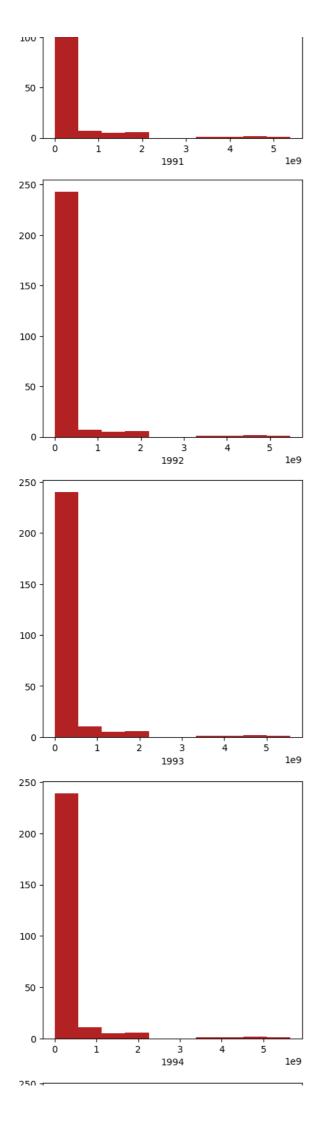


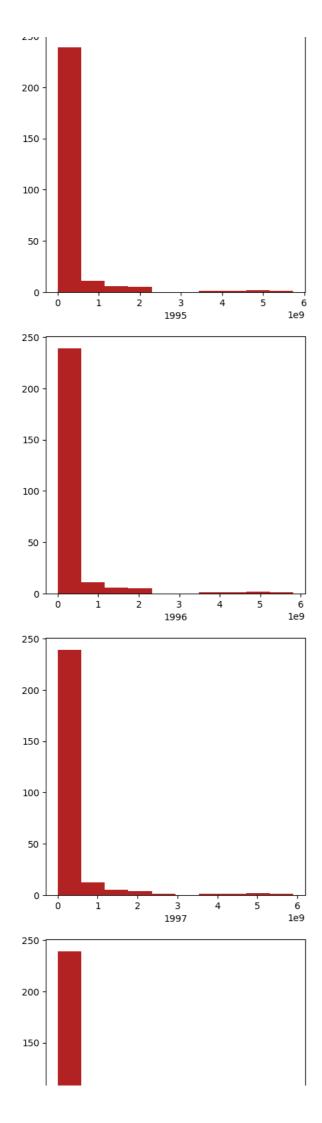


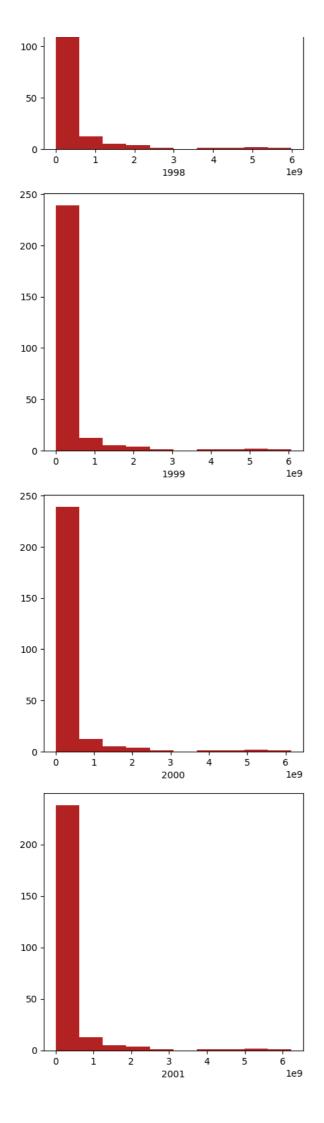


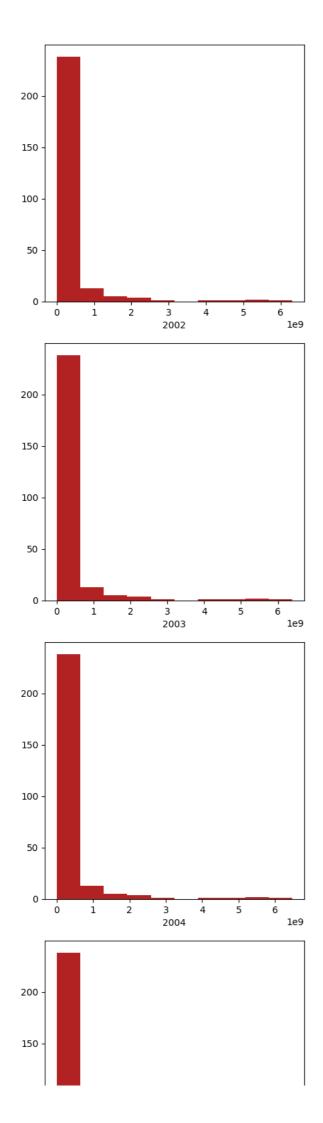


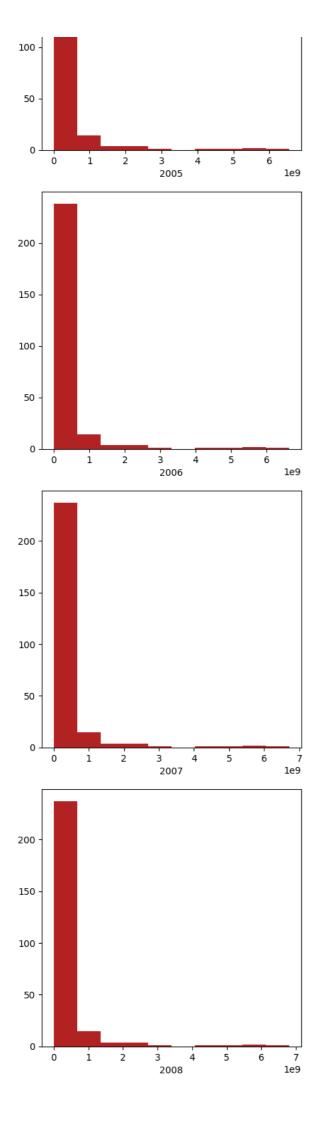


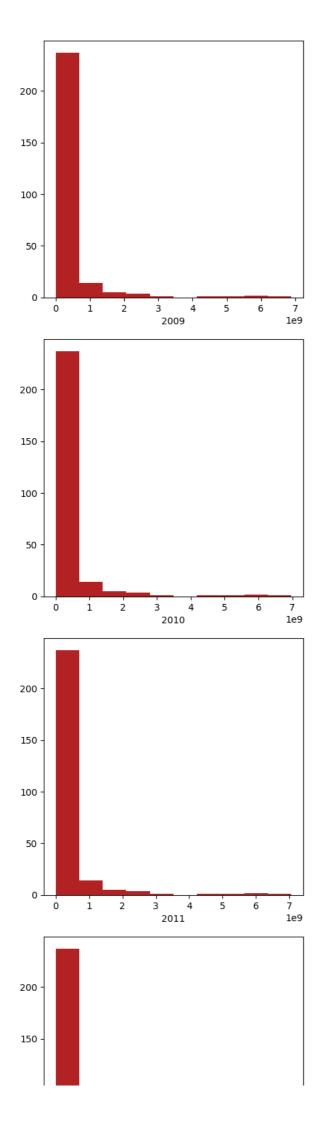


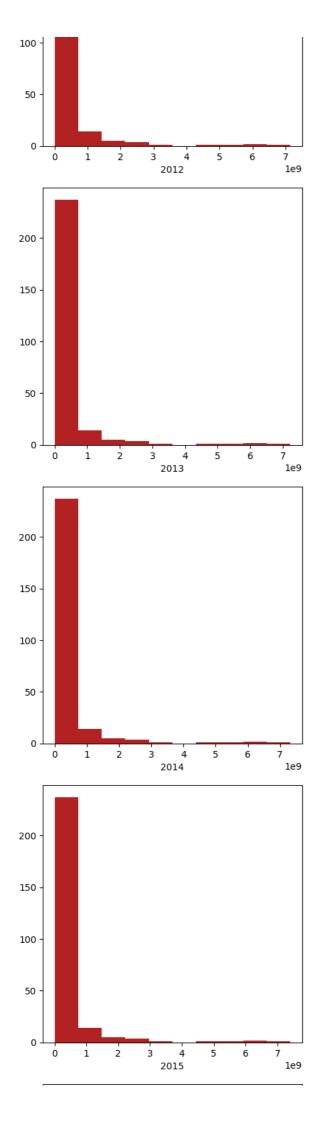


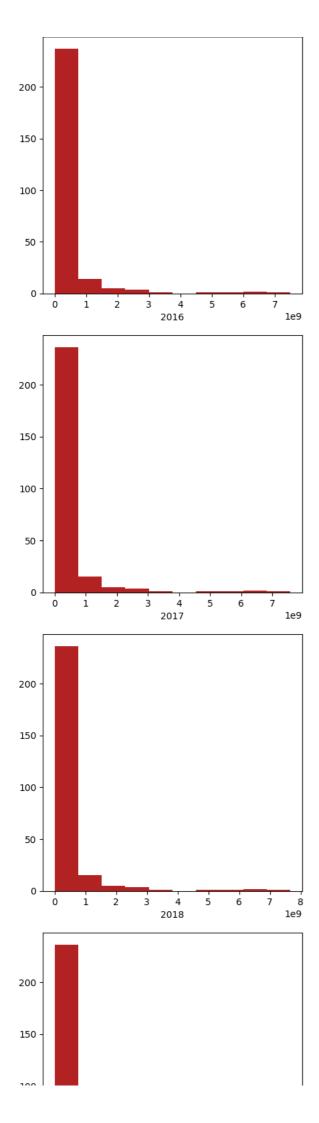


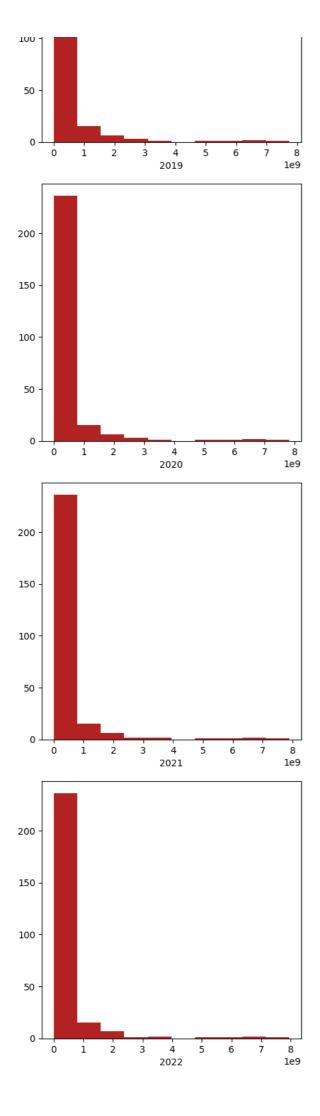








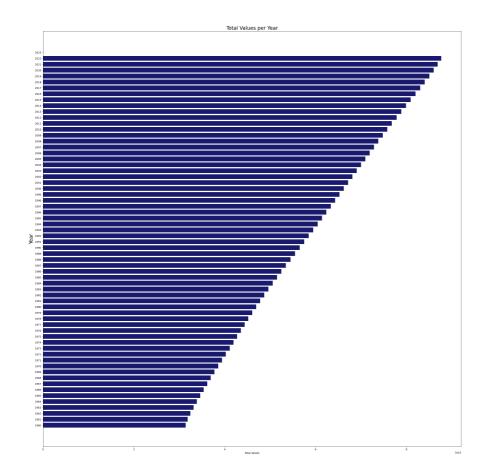




```
years = df.columns[1:]

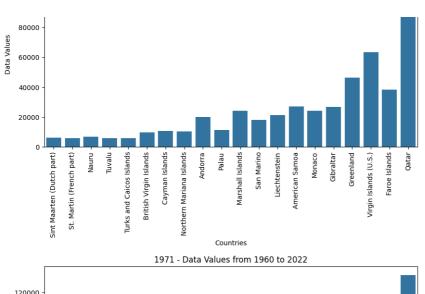
total_values = df[years].sum()

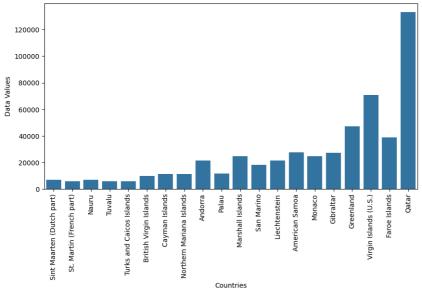
plt.figure(figsize=(30, 30))
plt.barh(years, total_values,color='#191970')
plt.xlabel('Total Values')
plt.ylabel('Year', size=20)
plt.title('Total Values per Year', size=20)
plt.show()
```

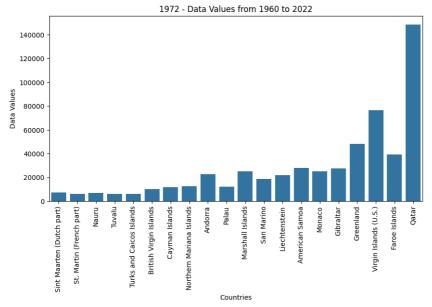


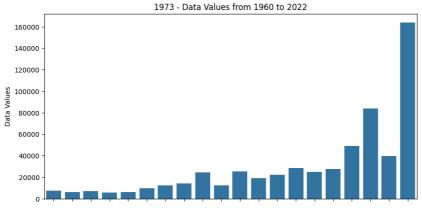
	Country Name	1960	1961	1962	1963	1964	1965	1966	1967
225	Sint Maarten (Dutch part)	2646.0	2888.0	3171.0	3481.0	3811.0	4161.0	4531.0	4930.0
147	St. Martin (French part)	4135.0	4258.0	4388.0	4524.0	4666.0	4832.0	5044.0	5294.0
179	Nauru	4582.0	4753.0	4950.0	5198.0	5484.0	5804.0	6021.0	6114.0
245	Tuvalu	5404.0	5436.0	5471.0	5503.0	5525.0	5548.0	5591.0	5657.0
228	Turks and Caicos Islands	5604.0	5625.0	5633.0	5634.0	5642.0	5650.0	5652.0	5662.0
255	British Virgin Islands	7850.0	7885.0	7902.0	7919.0	7949.0	8018.0	8139.0	8337.0
52	Cayman Islands	8473.0	8626.0	8799.0	8985.0	9172.0	9366.0	9566.0	9771.0
164	Northern Mariana Islands	8702.0	8965.0	9252.0	9561.0	9890.0	10229.0	10577.0	10720.0
6	Andorra	9443.0	10216.0	11014.0	11839.0	12690.0	13563.0	14546.0	15745.0
188	Palau	9446.0	9639.0	9851.0	10076.0	10318.0	10563.0	10813.0	10992.0
155	Marshall Islands	15374.0	15867.0	16387.0	16947.0	17537.0	18154.0	18794.0	19665.0
212	San Marino	15556.0	15895.0	16242.0	16583.0	16926.0	17273.0	17588.0	17907.0
137	Liechtenstein	16472.0	16834.0	17221.0	17625.0	18058.0	18500.0	18957.0	19467.0
11	American Samoa	20085.0	20626.0	21272.0	21949.0	22656.0	23391.0	24122.0	24848.0
149	Monaco	21797.0	21907.0	22106.0	22442.0	22766.0	23022.0	23198.0	23281.0
84	Gibraltar	21822.0	21907.0	22249.0	22796.0	23347.0	23910.0	24477.0	25047.0
91	Greenland	32500.0	33700.0	35000.0	36400.0	37600.0	39200.0	40500.0	41900.0
256	Virgin Islands (U.S.)	32500.0	34300.0	35000.0	39800.0	40800.0	43500.0	46200.0	49100.0
78	Faroe Islands	34154.0	34572.0	34963.0	35385.0	35841.0	36346.0	36825.0	37234.0
200	Qatar	36385.0	40111.0	45123.0	50950.0	57531.0	64843.0	73102.0	82517.0
20 rows × 65 columns									

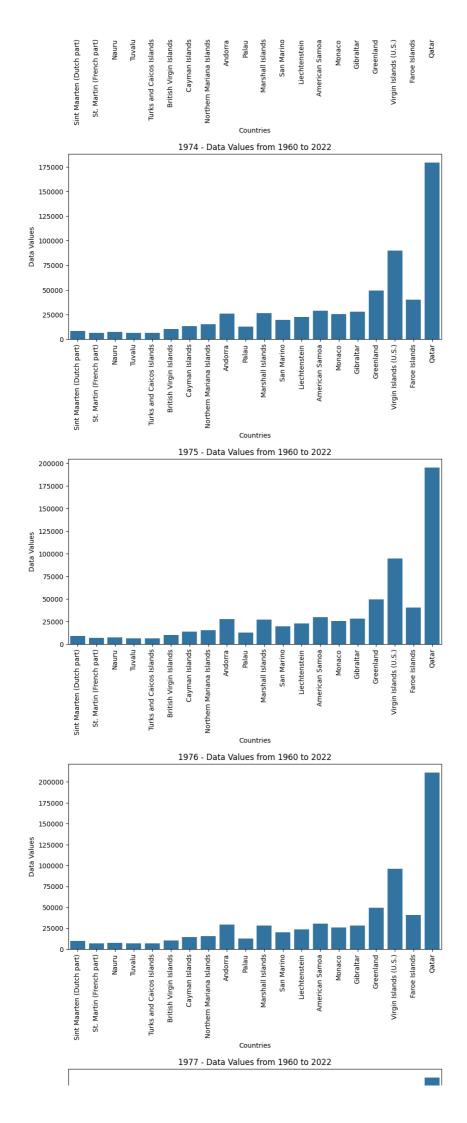
```
country_by_1960_t = country_by_1960.set_index('Country Name').T
for country_name, data_values in country_by_1960_t.iterrows():
    fig = plt.figure(figsize=(10, 5))
    sns.barplot(x=data_values.index, y=data_values.values)
    plt.xlabel('Countries')
    plt.ylabel('Data Values')
    plt.title(f"{country_name} - Data Values from 1960 to 2022")
    plt.xticks(rotation=90)
    plt.show()
```

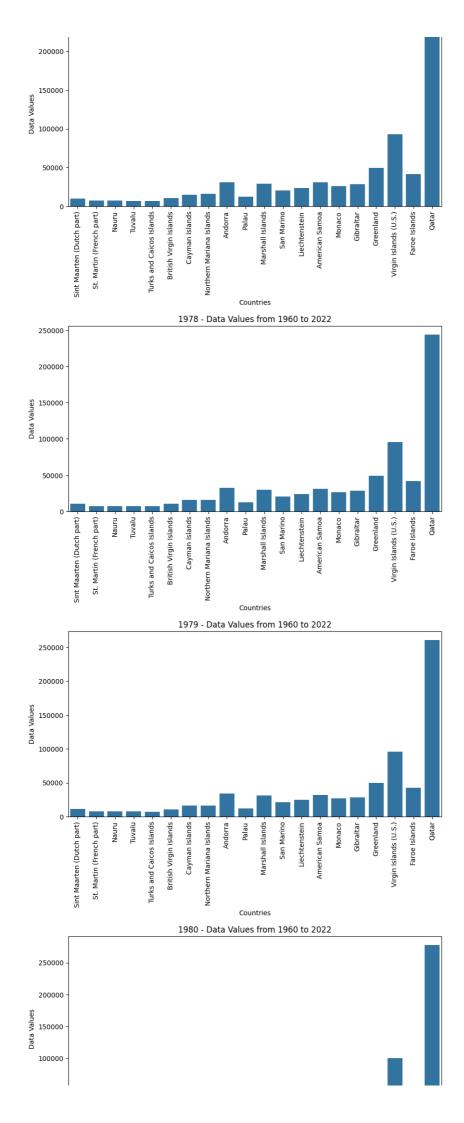


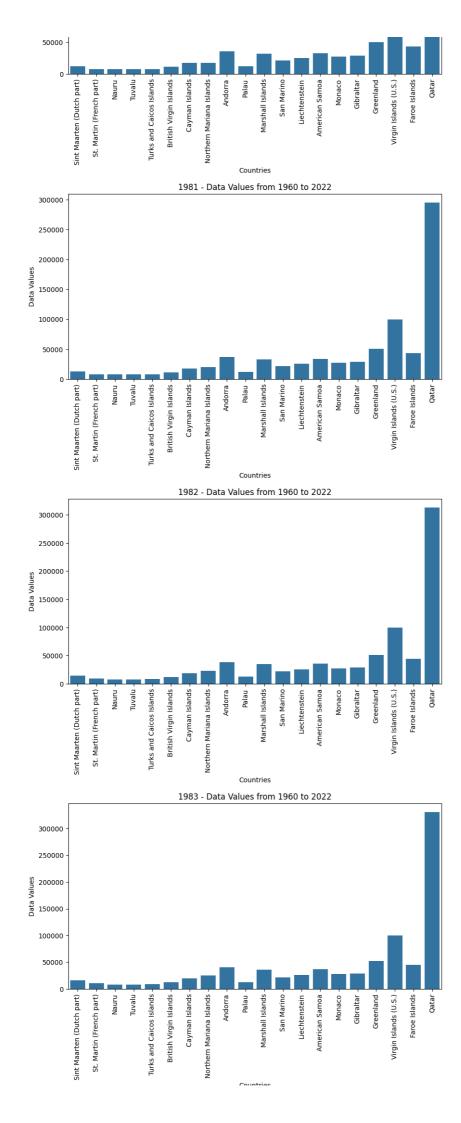




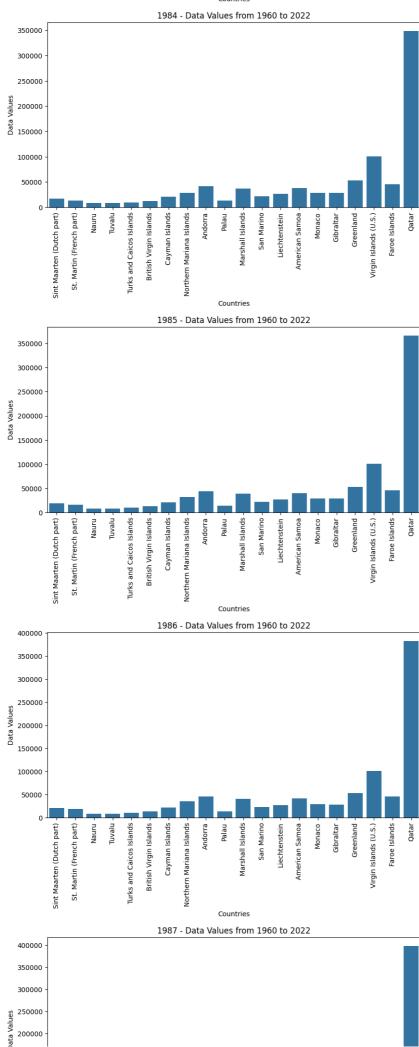


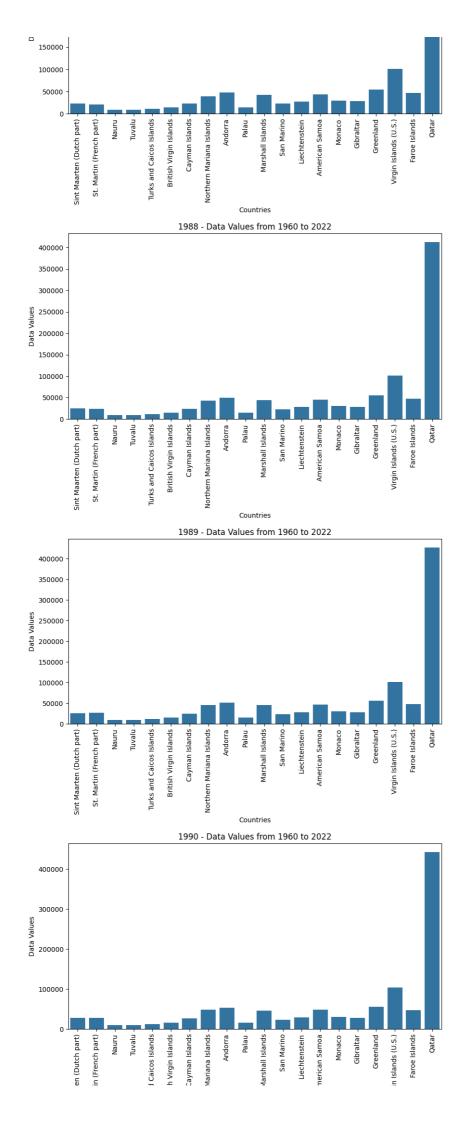


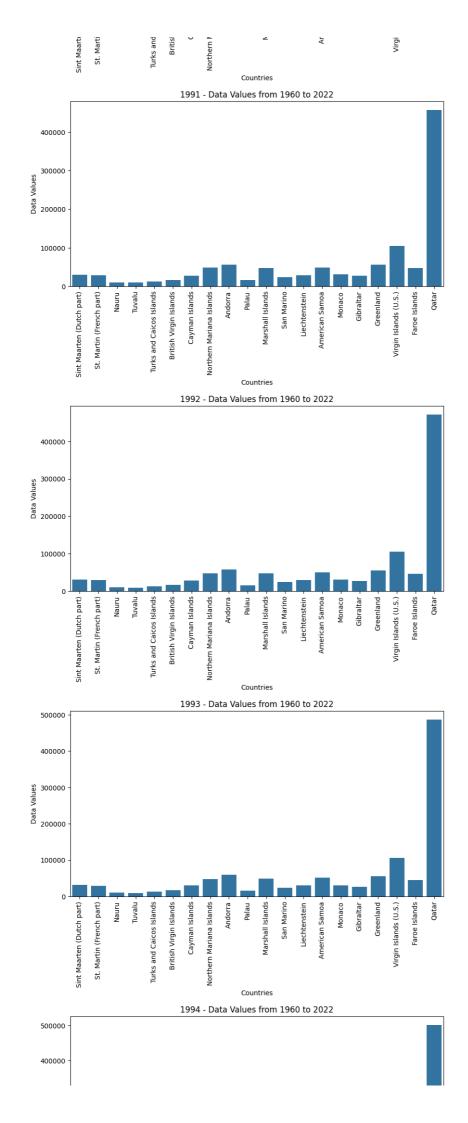


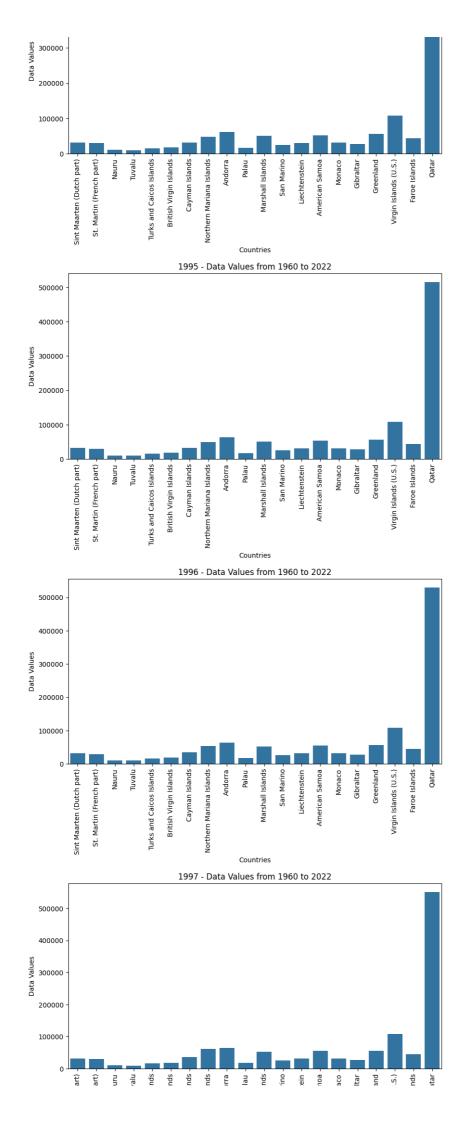


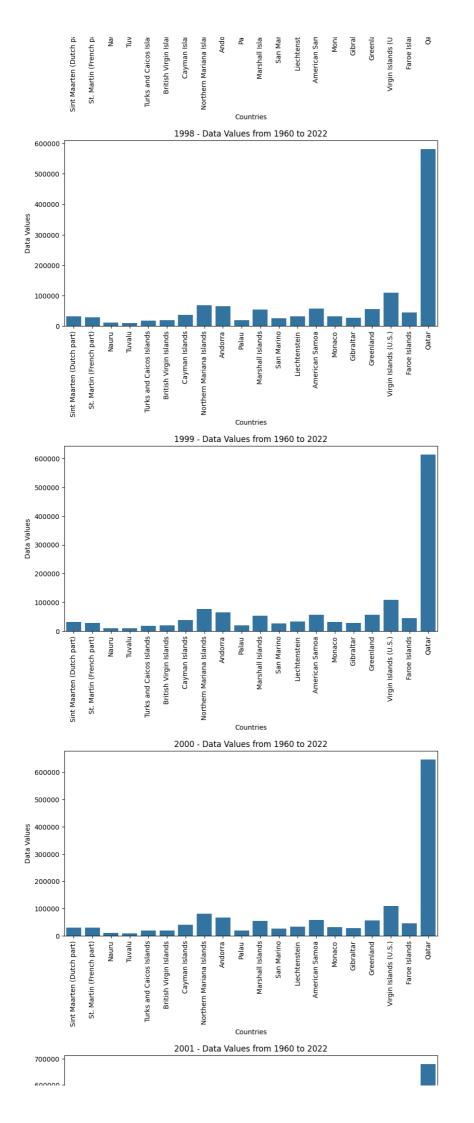
countries

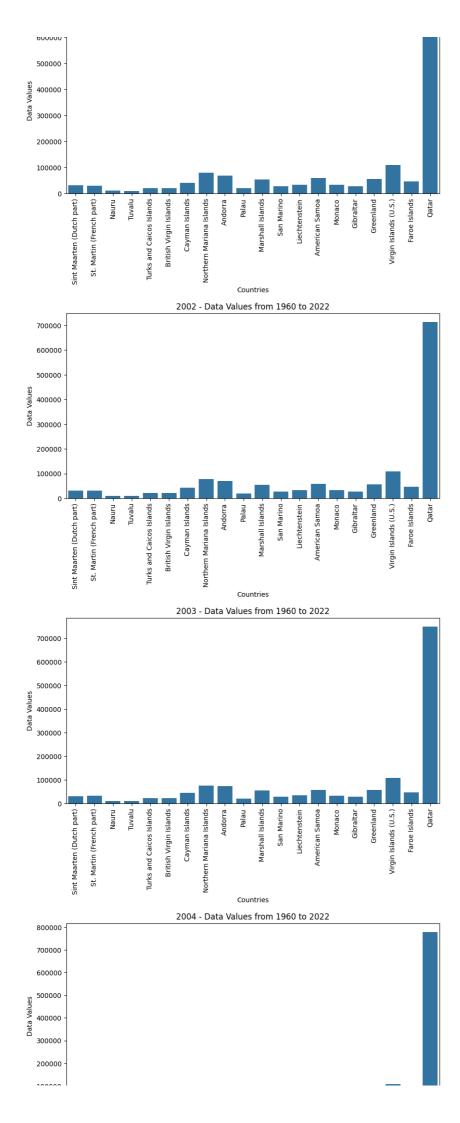


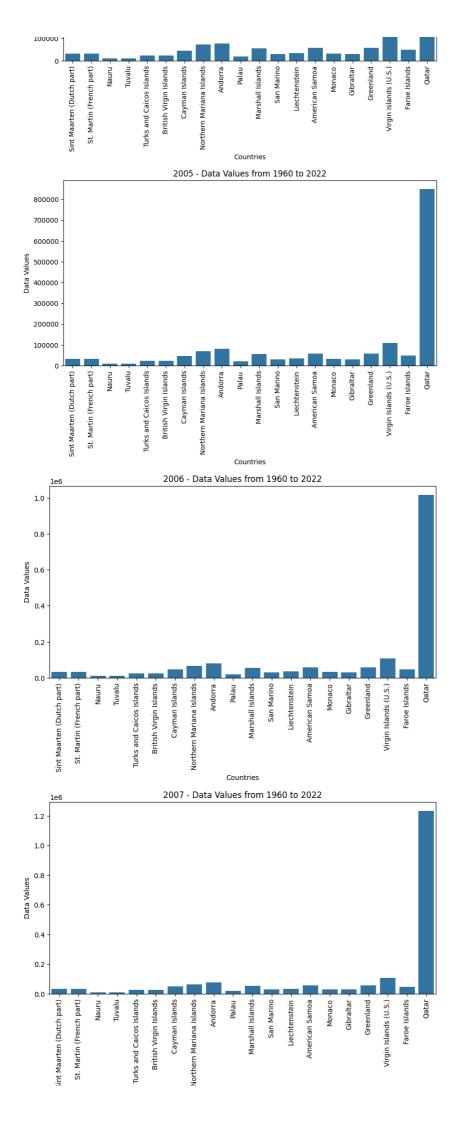


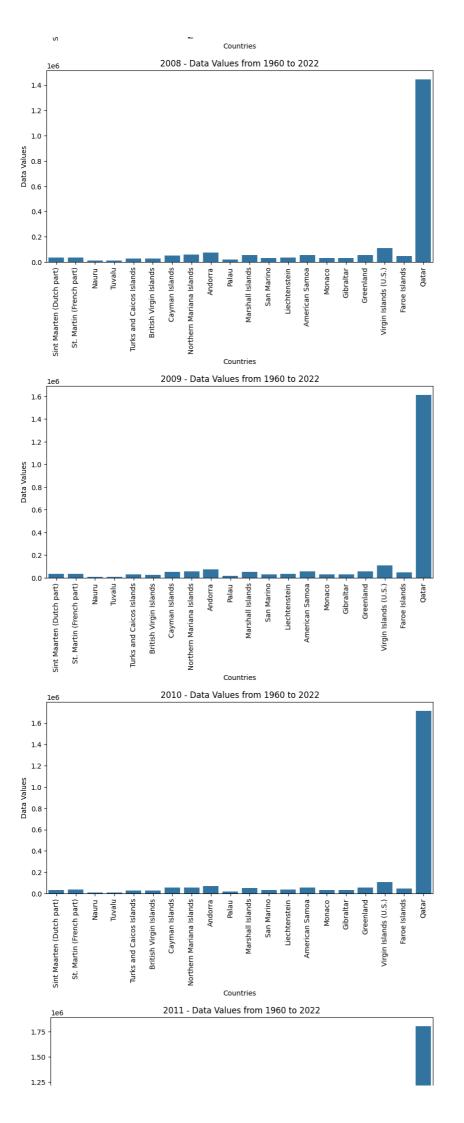


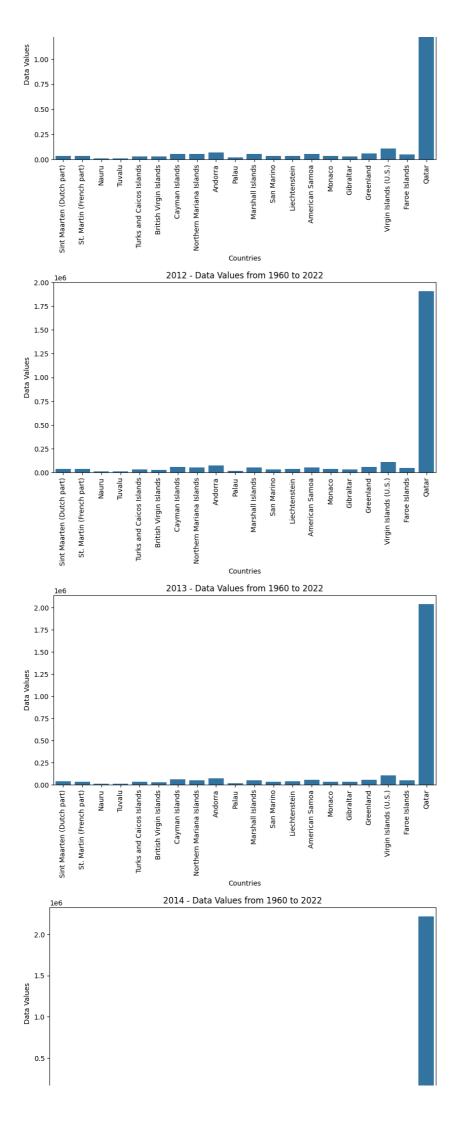


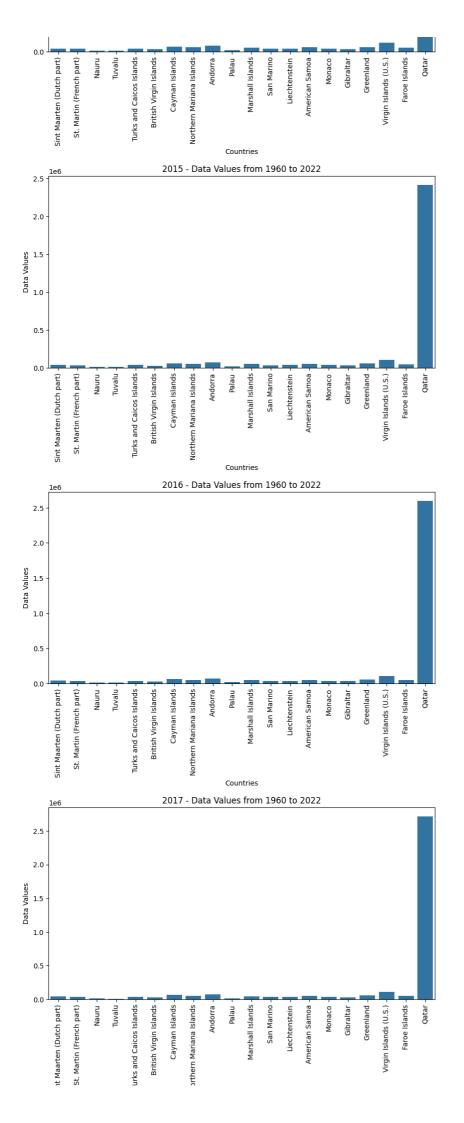


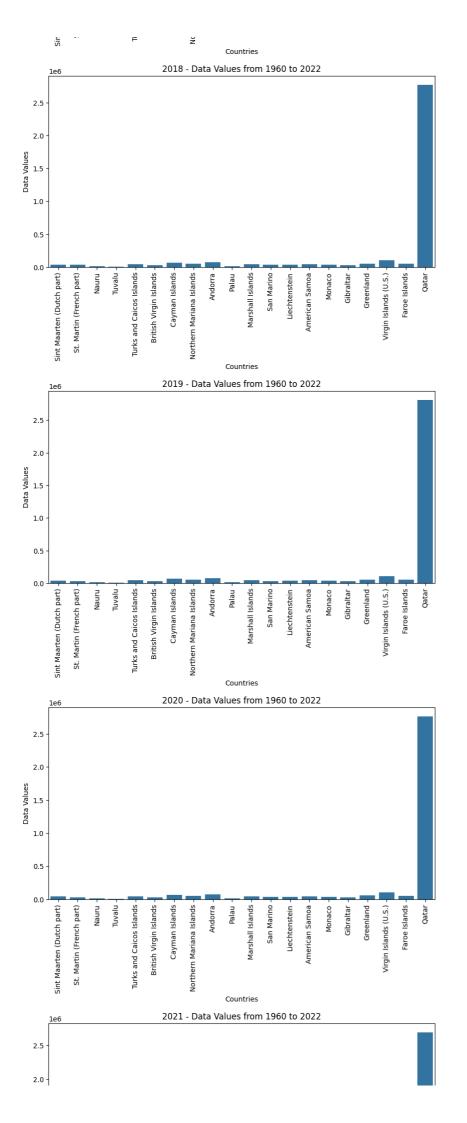






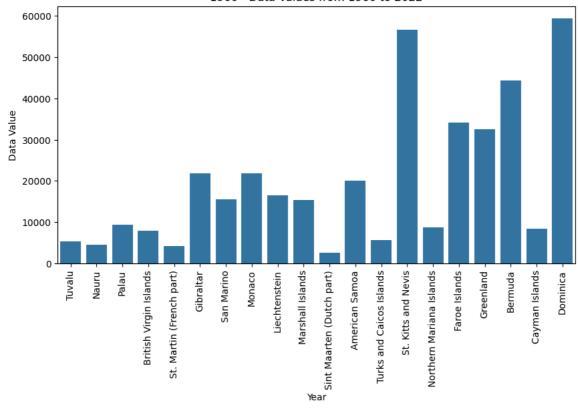


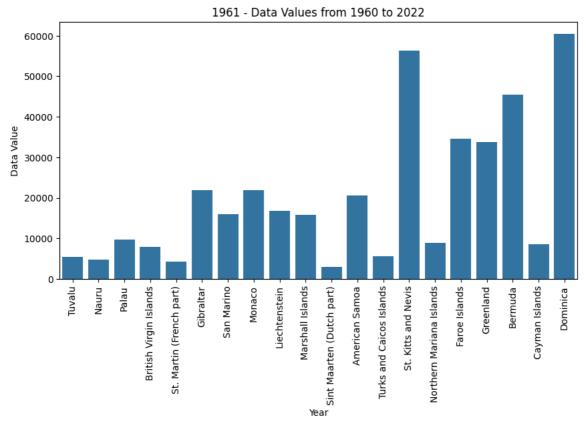


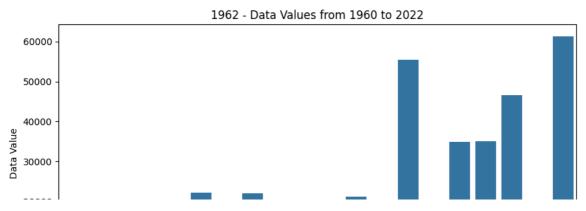


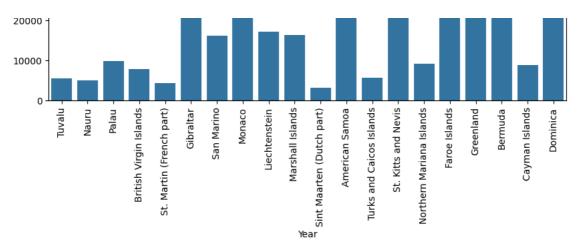
	Country Name	1960	1961	1962	1963	1964	1965	1966	1967
245	Tuvalu	5404.0	5436.0	5471.0	5503.0	5525.0	5548.0	5591.0	5657.0
179	Nauru	4582.0	4753.0	4950.0	5198.0	5484.0	5804.0	6021.0	6114.0
188	Palau	9446.0	9639.0	9851.0	10076.0	10318.0	10563.0	10813.0	10992.0
255	British Virgin Islands	7850.0	7885.0	7902.0	7919.0	7949.0	8018.0	8139.0	8337.0
147	St. Martin (French part)	4135.0	4258.0	4388.0	4524.0	4666.0	4832.0	5044.0	5294.0
84	Gibraltar	21822.0	21907.0	22249.0	22796.0	23347.0	23910.0	24477.0	25047.0
212	San Marino	15556.0	15895.0	16242.0	16583.0	16926.0	17273.0	17588.0	17907.0
149	Monaco	21797.0	21907.0	22106.0	22442.0	22766.0	23022.0	23198.0	23281.0
137	Liechtenstein	16472.0	16834.0	17221.0	17625.0	18058.0	18500.0	18957.0	19467.0
155	Marshall Islands	15374.0	15867.0	16387.0	16947.0	17537.0	18154.0	18794.0	19665.0
225	Sint Maarten (Dutch part)	2646.0	2888.0	3171.0	3481.0	3811.0	4161.0	4531.0	4930.0
11	American Samoa	20085.0	20626.0	21272.0	21949.0	22656.0	23391.0	24122.0	24848.0
228	Turks and Caicos Islands	5604.0	5625.0	5633.0	5634.0	5642.0	5650.0	5652.0	5662.0
125	St. Kitts and Nevis	56660.0	56247.0	55404.0	54391.0	53255.0	52016.0	50683.0	49269.0
164	Northern Mariana Islands	8702.0	8965.0	9252.0	9561.0	9890.0	10229.0	10577.0	10720.0
78	Faroe Islands	34154.0	34572.0	34963.0	35385.0	35841.0	36346.0	36825.0	37234.0
91	Greenland	32500.0	33700.0	35000.0	36400.0	37600.0	39200.0	40500.0	41900.0
27	Bermuda	44400.0	45500.0	46600.0	47700.0	48900.0	50100.0	51000.0	52000.0
52	Cayman Islands	8473.0	8626.0	8799.0	8985.0	9172.0	9366.0	9566.0	9771.0
57	Dominica	59379.0	60395.0	61224.0	62031.0	62843.0	63744.0	64728.0	65760.0
20 rov	20 rows × 65 columns								

```
country_by_2022_t = country_by_2022.set_index('Country Name').T
for country_name, data_values in country_by_2022_t.iterrows():
    fig = plt.figure(figsize=(10, 5))
    sns.barplot(x=data_values.index, y=data_values.values)
    plt.xlabel('Year')
    plt.ylabel('Data Value')
    plt.title(f"{country_name} - Data Values from 1960 to 2022")
    plt.xticks(rotation=90)
    plt.show()
```

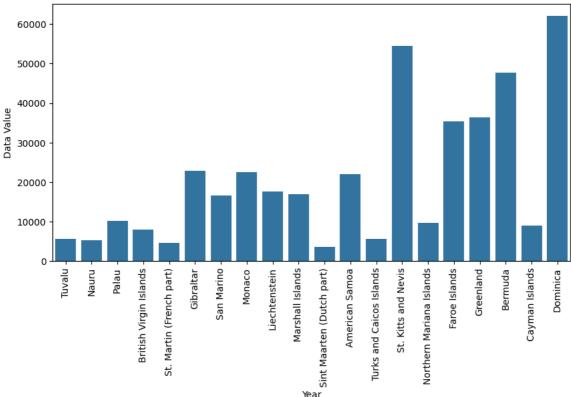




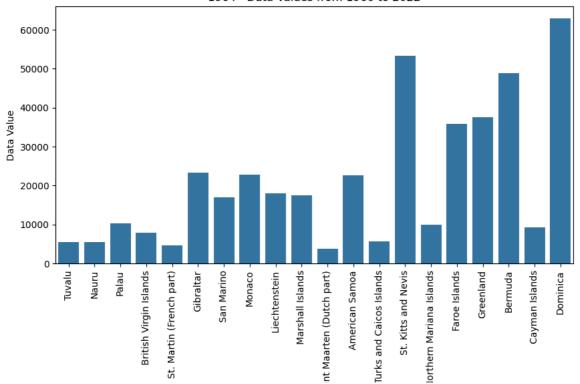




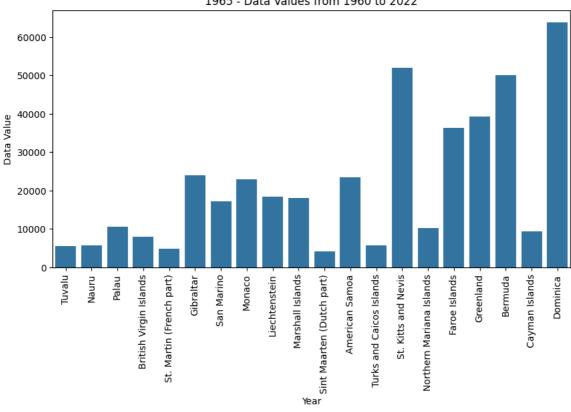
1963 - Data Values from 1960 to 2022



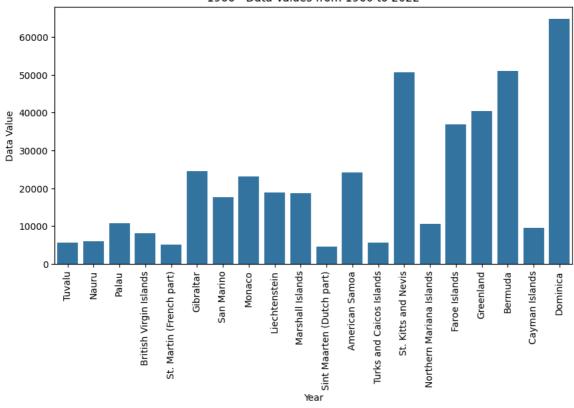
1964 - Data Values from 1960 to 2022



## 1965 - Data Values from 1960 to 2022



## 1966 - Data Values from 1960 to 2022



## 1967 - Data Values from 1960 to 2022

