In [3]: import numpy as np
 import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt

In [35]: df1=pd.read_csv('APIdataset.csv')
 df1

Out[35]:

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1!
0	Aruba	ABW	Population, total	SP.POP.TOTL	5.460800e+04	5.581100e+04	5.668200e·
1	Africa Eastern and Southern	AFE	Population, total	SP.POP.TOTL	1.306926e+08	1.341692e+08	1.378356e·
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8.622466e+06	8.790140e+06	8.969047e-
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	9.725629e+07	9.931403e+07	1.014450e·
4	Angola	AGO	Population, total	SP.POP.TOTL	5.357195e+06	5.441333e+06	5.521400e·
5	Albania	ALB	Population, total	SP.POP.TOTL	1.608800e+06	1.659800e+06	1.711319e· ▼

In [36]: df1.head()

Out[36]:

	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	141
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8622466.0	8790140.0	8969047.0	9
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	97256290.0	99314028.0	101445032.0	103
4	Angola	AGO	Population, total	SP.POP.TOTL	5357195.0	5441333.0	5521400.0	5

5 rows × 68 columns

```
In [37]: df1.tail()
```

Out[37]:

	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	1022
262	Yemen, Rep.	YEM	Population, total	SP.POP.TOTL	5542459.0	5646668.0	5753386.0	5860
263	South Africa	ZAF	Population, total	SP.POP.TOTL	16520441.0	16989464.0	17503133.0	18042
264	Zambia	ZMB	Population, total	SP.POP.TOTL	3119430.0	3219451.0	3323427.0	3431
265	Zimbabwe	ZWE	Population, total	SP.POP.TOTL	3806310.0	3925952.0	4049778.0	4177

5 rows × 68 columns

```
In [38]: df1.shape
```

In [38]: uti.Snape

Out[38]: (266, 68)

```
In [39]: df1.columns
```

```
Out[39]: Index(['Country Name', 'Country Code', 'Indicator Name', 'Indicator Code',
                 '1960', '1961', '1962', '1963', '1964', '1965', '1966', '1967', '19
         68',
                '1969', '1970', '1971', '1972', '1973', '1974', '1975', '1976', '19
         77',
                '1978', '1979', '1980', '1981', '1982', '1983', '1984', '1985', '19
         86',
                '1987', '1988', '1989', '1990', '1991', '1992', '1993', '1994', '19
         95',
                '1996', '1997', '1998', '1999', '2000', '2001', '2002', '2003', '20
         04',
                '2005', '2006', '2007', '2008', '2009', '2010', '2011', '2012', '20
         13',
                '2014', '2015', '2016', '2017', '2018', '2019', '2020', '2021', '20
         22',
                '2023'],
               dtype='object')
```

In [12]: df1.dtypes

,				3
Out[12]:	Data Soul	rce		object
	World De	velopment	Indicators	object
	Unnamed:	2		object
	Unnamed:	3		object
	Unnamed:	4		float64
	Unnamed:	5		float64
	Unnamed:	6		float64
	Unnamed:	7		float64
	Unnamed:	8		float64
	Unnamed: Unnamed:	9		float64
	Unnamed:	10 11		float64 float64
	Unnamed:	12		float64
	Unnamed:	13		float64
	Unnamed:	14		float64
	Unnamed:	15		float64
	Unnamed:	16		float64
	Unnamed:	17		float64
	Unnamed:	18		float64
	Unnamed:	19		float64
	Unnamed:	20		float64
	Unnamed:	21		float64
	Unnamed:	22		float64
	Unnamed:	23		float64
	Unnamed:	24		float64
	Unnamed:	25		float64
	Unnamed:	26		float64
	Unnamed:	27		float64
	Unnamed:	28		float64
	Unnamed:	29		float64
				• • •
	Unnamed:	38		float64
	Unnamed:	39		float64
	Unnamed:	40		float64
	Unnamed:	41		float64
	Unnamed:	42		float64
	Unnamed:	43		float64
	Unnamed:	44		float64
	Unnamed:	45		float64
	Unnamed:	46		float64
	Unnamed:	47		float64
	Unnamed:	48		float64
	Unnamed:	49		float64
	Unnamed:	50 51		float64
	Unnamed:	51 52		float64
	Unnamed: Unnamed:	52 53		float64 float64
	Unnamed:	55 54		float64
	Unnamed:	55		float64
	Unnamed:	56		float64
	Unnamed:	57		float64
	Unnamed:	58		float64
	Unnamed:	59		float64
	Unnamed:	60		float64
	Unnamed:	61		float64
	Unnamed:	62		float64
	Unnamed:	63		float64
	Unnamed:	64		float64
	Unnamed:	65		float64
	Unnamed:	66		float64

Unnamed: 67 float64

Length: 68, dtype: object

In [40]: df1.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 266 entries, 0 to 265 Data columns (total 68 columns): Country Name 266 non-null object Country Code 266 non-null object Indicator Name 266 non-null object Indicator Code 266 non-null object 1960 264 non-null float64 264 non-null float64 1961 1962 264 non-null float64 264 non-null float64 1963 1964 264 non-null float64 264 non-null float64 1965 264 non-null float64 1966 264 non-null float64 1967 264 non-null float64 1968 264 non-null float64 1969 1970 264 non-null float64 264 non-null float64 1971 1972 264 non-null float64 1973 264 non-null float64 1974 264 non-null float64 1975 264 non-null float64 264 non-null float64 1976 264 non-null float64 1977 264 non-null float64 1978 1979 264 non-null float64 264 non-null float64 1980 264 non-null float64 1981 264 non-null float64 1982 1983 264 non-null float64 1984 264 non-null float64 264 non-null float64 1985 1986 264 non-null float64 264 non-null float64 1987 264 non-null float64 1988 264 non-null float64 1989 1990 265 non-null float64 1991 265 non-null float64 1992 265 non-null float64 1993 265 non-null float64 1994 265 non-null float64 1995 265 non-null float64 265 non-null float64 1996 1997 265 non-null float64 265 non-null float64 1998 265 non-null float64 1999 265 non-null float64 2000 265 non-null float64 2001 265 non-null float64 2002 2003 265 non-null float64 265 non-null float64 2004 265 non-null float64 2005 2006 265 non-null float64 265 non-null float64 2007 2008 265 non-null float64 265 non-null float64 2009 265 non-null float64 2010 265 non-null float64 2011 2012 265 non-null float64 265 non-null float64 2013

```
2014
                  265 non-null float64
                  265 non-null float64
2015
2016
                  265 non-null float64
                  265 non-null float64
2017
                  265 non-null float64
2018
2019
                  265 non-null float64
                  265 non-null float64
2020
2021
                  265 non-null float64
2022
                  265 non-null float64
2023
                  0 non-null float64
```

dtypes: float64(64), object(4)

memory usage: 141.4+ KB

In [41]: df1.describe()

Out[41]:

	1960	1961	1962	1963	1964	1965
count	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+02	2.640000e+02
mean	1.172860e+08	1.188956e+08	1.210661e+08	1.237484e+08	1.264530e+08	1.291965e+08
std	3.695500e+08	3.740958e+08	3.808121e+08	3.895098e+08	3.982497e+08	4.071209e+08
min	2.646000e+03	2.888000e+03	3.171000e+03	3.481000e+03	3.811000e+03	4.161000e+03
25%	5.132212e+05	5.231345e+05	5.337595e+05	5.449288e+05	5.566630e+05	5.651150e+05
50%	3.757486e+06	3.887144e+06	4.023896e+06	4.139356e+06	4.224612e+06	4.277636e+06
75%	2.670606e+07	2.748694e+07	2.830289e+07	2.914708e+07	3.001684e+07	3.084892e+07
max	3.031474e+09	3.072422e+09	3.126850e+09	3.193429e+09	3.260442e+09	3.328209e+09

8 rows × 64 columns

In [42]: df1.duplicated().sum()

Out[42]: 0

In [43]: df1.isna().sum().any()

Out[43]: True

In [44]: df1=df1.fillna(method="ffill")
 df1.head()

Out[44]:

	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	141
2	Afghanistan	AFG	Population, total	SP.POP.TOTL	8622466.0	8790140.0	8969047.0	9
3	Africa Western and Central	AFW	Population, total	SP.POP.TOTL	97256290.0	99314028.0	101445032.0	103
4	Angola	AGO	Population, total	SP.POP.TOTL	5357195.0	5441333.0	5521400.0	5

5 rows × 68 columns

In [45]: df1.isna().sum().any()

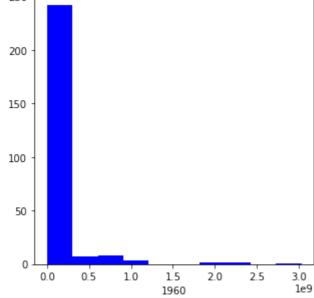
Out[45]: True

In [46]: df1['Country Name'].unique()

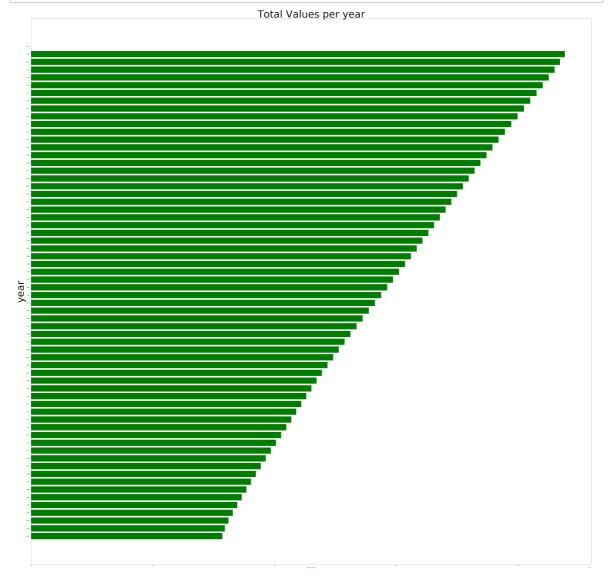
Out[46]: array(['Aruba', 'Africa Eastern and Southern', 'Afghanistan', 'Africa Western and Central', 'Angola', 'Albania', 'Andorra', 'Arab World', 'United Arab Emirates', 'Argentina', 'Armenia', 'American Samoa', 'Antigua and Barbuda', 'Australia', 'Austria', 'Azerbaijan', 'Burundi', 'Belgium', 'Benin', 'Burkina Faso', 'Bangladesh', 'Bulgaria', 'Bahrain', 'Bahamas, The', 'Bosnia and Herzegovina', 'Belarus', 'Belize', 'Bermuda', 'Bolivia', 'Brazil', 'Barbados', 'Brunei Darussalam', 'Bhutan', 'Botswana', 'Central African Republic', 'Canada', 'Central Europe and the Baltics', 'Switzerland', 'Channel Islands', 'Chile', 'China', "Cote d'Ivoire", 'Cameroon', 'Congo, Dem. Rep.', 'Congo, Rep.', 'Colombia', 'Comoros', 'Cabo Verde', 'Costa Rica', 'Caribbean small states', 'Cuba', 'Curacao', 'Cayman Islands', 'Cyprus', 'Czechia', 'Germany', 'Djibouti', 'Dominica', 'Denmark', 'Dominican Republic', 'Algeria', 'East Asia & Pacific (excluding high income)', 'Early-demographic dividend', 'East Asia & Pacific', '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' '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)
In [47]: df1['Indicator Code'].unique()
Out[47]: array(['SP.POP.TOTL'], dtype=object)
          df1.drop(['Indicator Name','Indicator Code','Country Code'],axis=1,inplace=T
In [50]: df1.columns
Out[50]: Index(['Country Name', '1960', '1961', '1962', '1963', '1964', '1965', '19
          66',
                  '1967', '1968', '1969', '1970', '1971', '1972', '1973', '1974', '19
          75',
                  '1976', '1977', '1978', '1979', '1980', '1981', '1982', '1983', '19
          84',
                  '1985', '1986', '1987', '1988', '1989', '1990', '1991', '1992', '19
          93',
                  '1994', '1995', '1996', '1997', '1998', '1999', '2000', '2001', '20
          02',
                  '2003', '2004', '2005', '2006', '2007', '2008', '2009', '2010', '20
          11',
                  '2012', '2013', '2014', '2015', '2016', '2017', '2018', '2019', '20
          20',
                  '2021', '2022', '2023'],
                 dtype='object')
In [53]: cols=['1960', '1961', '1962', '1963', '1964', '1965', '1966',
                   '1967', '1968', '1969', '1970', '1971', '1972', '1973', '1974', '1975
                   '1976', '1977', '1978', '1979', '1980', '1981', '1982', '1983', '1984
                  '1985', '1986', '1987', '1988', '1989', '1990', '1991', '1992', '1993'
'1994', '1995', '1996', '1997', '1998', '1999', '2000', '2001', '2002'
'2003', '2004', '2005', '2006', '2007', '2008', '2009', '2010', '2011'
'2012', '2013', '2014', '2015', '2016', '2017', '2018', '2019', '2020'
                   '2021', '2022', '2023']
```

```
In [56]: for i in cols:
    fig=plt.figure(figsize=(5,5))
    plt.hist(df1[i],color='blue',bins=10)
    plt.xlabel(i)
    plt.show()
```



```
In [60]: years=df1.columns[1:]
    total_values=df1[years].sum()
    plt.figure(figsize=(70,70))
    plt.barh(years,total_values,color='green')
    plt.xlabel("Total Values")
    plt.ylabel('year',size=70)
    plt.title("Total Values per year",size=70)
    plt.show()
```



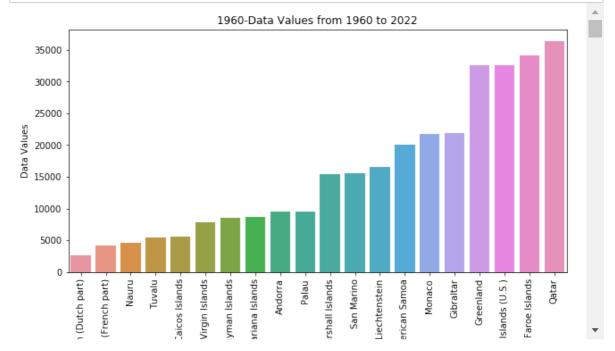
In [61]: country_by_1960=df1.sort_values(by='1960').head(20)
country_by_1960

Out[61]:

	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	53
147	St. Martin (French part)	4135.0	4258.0	4388.0	4524.0	4666.0	4832.0	5044.0	5294.0	54
179	Nauru	4582.0	4753.0	4950.0	5198.0	5484.0	5804.0	6021.0	6114.0	62
245	Tuvalu	5404.0	5436.0	5471.0	5503.0	5525.0	5548.0	5591.0	5657.0	57
228	Turks and Caicos Islands	5604.0	5625.0	5633.0	5634.0	5642.0	5650.0	5652.0	5662.0	56
255	British Virgin Islands	7850.0	7885.0	7902.0	7919.0	7949.0	8018.0	8139.0	8337.0	86
52	Cayman Islands	8473.0	8626.0	8799.0	8985.0	9172.0	9366.0	9566.0	9771.0	99
164	Northern Mariana Islands	8702.0	8965.0	9252.0	9561.0	9890.0	10229.0	10577.0	10720.0	104
6	Andorra	9443.0	10216.0	11014.0	11839.0	12690.0	13563.0	14546.0	15745.0	170
188	Palau	9446.0	9639.0	9851.0	10076.0	10318.0	10563.0	10813.0	10992.0	110
155	Marshall Islands	15374.0	15867.0	16387.0	16947.0	17537.0	18154.0	18794.0	19665.0	210
212	San Marino	15556.0	15895.0	16242.0	16583.0	16926.0	17273.0	17588.0	17907.0	182
137	Liechtenstein	16472.0	16834.0	17221.0	17625.0	18058.0	18500.0	18957.0	19467.0	200
11	American Samoa	20085.0	20626.0	21272.0	21949.0	22656.0	23391.0	24122.0	24848.0	256
149	Monaco	21797.0	21907.0	22106.0	22442.0	22766.0	23022.0	23198.0	23281.0	234
84	Gibraltar	21822.0	21907.0	22249.0	22796.0	23347.0	23910.0	24477.0	25047.0	256
91	Greenland	32500.0	33700.0	35000.0	36400.0	37600.0	39200.0	40500.0	41900.0	434
256	Virgin Islands (U.S.)	32500.0	34300.0	35000.0	39800.0	40800.0	43500.0	46200.0	49100.0	557
78	Faroe Islands	34154.0	34572.0	34963.0	35385.0	35841.0	36346.0	36825.0	37234.0	376
200	Qatar	36385.0	40111.0	45123.0	50950.0	57531.0	64843.0	73102.0	82517.0	930

20 rows × 65 columns

```
In [64]: 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()
```



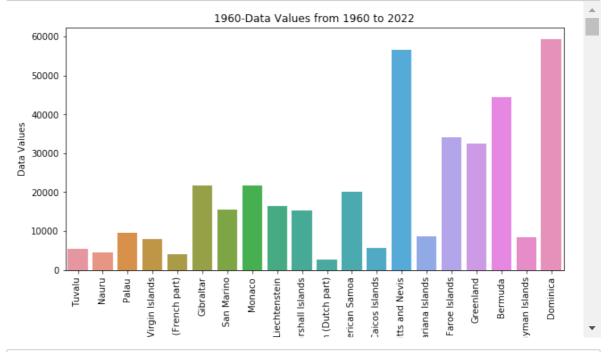
In [66]: country_by_2022=df1.sort_values(by='2022').head(20)
country_by_2022

Out[66]:

	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	57
179	Nauru	4582.0	4753.0	4950.0	5198.0	5484.0	5804.0	6021.0	6114.0	62
188	Palau	9446.0	9639.0	9851.0	10076.0	10318.0	10563.0	10813.0	10992.0	110
255	British Virgin Islands	7850.0	7885.0	7902.0	7919.0	7949.0	8018.0	8139.0	8337.0	86
147	St. Martin (French part)	4135.0	4258.0	4388.0	4524.0	4666.0	4832.0	5044.0	5294.0	54
84	Gibraltar	21822.0	21907.0	22249.0	22796.0	23347.0	23910.0	24477.0	25047.0	256
212	San Marino	15556.0	15895.0	16242.0	16583.0	16926.0	17273.0	17588.0	17907.0	182
149	Monaco	21797.0	21907.0	22106.0	22442.0	22766.0	23022.0	23198.0	23281.0	234
137	Liechtenstein	16472.0	16834.0	17221.0	17625.0	18058.0	18500.0	18957.0	19467.0	200
155	Marshall Islands	15374.0	15867.0	16387.0	16947.0	17537.0	18154.0	18794.0	19665.0	210
225	Sint Maarten (Dutch part)	2646.0	2888.0	3171.0	3481.0	3811.0	4161.0	4531.0	4930.0	53
11	American Samoa	20085.0	20626.0	21272.0	21949.0	22656.0	23391.0	24122.0	24848.0	256
228	Turks and Caicos Islands	5604.0	5625.0	5633.0	5634.0	5642.0	5650.0	5652.0	5662.0	56
125	St. Kitts and Nevis	56660.0	56247.0	55404.0	54391.0	53255.0	52016.0	50683.0	49269.0	477
164	Northern Mariana Islands	8702.0	8965.0	9252.0	9561.0	9890.0	10229.0	10577.0	10720.0	104
78	Faroe Islands	34154.0	34572.0	34963.0	35385.0	35841.0	36346.0	36825.0	37234.0	376
91	Greenland	32500.0	33700.0	35000.0	36400.0	37600.0	39200.0	40500.0	41900.0	434
27	Bermuda	44400.0	45500.0	46600.0	47700.0	48900.0	50100.0	51000.0	52000.0	530
52	Cayman Islands	8473.0	8626.0	8799.0	8985.0	9172.0	9366.0	9566.0	9771.0	99
57	Dominica	59379.0	60395.0	61224.0	62031.0	62843.0	63744.0	64728.0	65760.0	668

20 rows × 65 columns

```
In [67]:
    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 Values')
        plt.title(f"{country_name}-Data Values from 1960 to 2022")
        plt.xticks(rotation=90)
        plt.show()
```



In []: