

Climate-Change-Indicators

March 15, 2024

1 Project Summary :

- The Climate Change Indicators Analysis project aims to assess and analyze various indicators of climate change across different regions and time periods.
- By examining key metrics such as temperature change along varying years and other relevant factors, the project seeks to provide insights into the ongoing impacts of climate change on the environment.
- The **Average surface temperature on Earth** is approximately **15 degrees Celsius**, according to NASA and we compared the Difference of Mean temperature Changes, Max and Min Temperature Changes with Average surface temperature on Earth.
- This project involved exploring and cleaning a dataset to prepare it for analysis. The data exploration process involved identifying and understanding the characteristics of the data, such as the data types, missing values, and distributions of values
- The data cleaning process involved identifying and addressing any issues or inconsistencies in the data, such as errors, missing values, or duplicate records
- Through this process, we were able to identify and fix any issues with the data, and ensure that it was ready for further analysis
- The clean and prepared data can now be used to answer specific research.
- Once the data has been cleaned and prepared, now begin exploring and summarizing it with describe the data using data manipulation using Pandas to explore and understand patterns in Climate Change data.
- We created various charts to visualize the data, and wrote observations and insights below each one to help us better understand the data and identify useful insights and patterns.
- The observations and insights we identified through this process will be useful for future analysis and decision-making related to Climate Change

2 Problem Statement :

- Analyze the distribution of temperature changes across countries over the years.
- Identify countries with the highest and lowest average temperature changes.
- Calculate the total temperature change for each country across all years.
- Visualize the data using appropriate plots to facilitate better understanding and interpretation.
- Identification of countries with the highest and lowest temperature changes.
- Visualization of temperature change trends using histograms, bar plots, or other suitable visualizations.
- This Dataset can provide valuable insights into how temperature changes vary across countries and regions, which can inform decision-makers, researchers, and policymakers in addressing the future challenges posed by climate change.

3.0.1 Importing necessary Libraries and Load Climate Change Dataset

```
[1]: import pandas as pd

Climate_df=pd.read_csv(r C:\Users\Admin\Downloads\climate_change_indicators.
↳CSV )
Climate_df
```

```
[1]:
```

	ObjectId	Country	ISO2	ISO3	\
0	1	Afghanistan, Islamic Rep. of	AF	AFG	
1	2	Albania	AL	ALB	
2	3	Algeria	DZ	DZA	
3	4	American Samoa	AS	ASM	
4	5	Andorra, Principality of	AD	AND	
..
220	221	Western Sahara	EH	ESH	221 222
		World	NaN	WLD	
222	223	Yemen, Rep. of	YE	YEM	
223	224	Zambia	ZM	ZMB	
224	225	Zimbabwe	ZW	ZWE	

	Indicator	Unit	\
0	Temperature change with respect to a baseline ...	Degree Celsius	
1	Temperature change with respect to a baseline ...	Degree Celsius	
2	Temperature change with respect to a baseline ...	Degree Celsius	
3	Temperature change with respect to a baseline ...	Degree Celsius	
4	Temperature change with respect to a baseline ...	Degree Celsius	
..	
220	Temperature change with respect to a baseline ...	Degree Celsius	
221	Temperature change with respect to a baseline ...	Degree Celsius	
222	Temperature change with respect to a baseline ...	Degree Celsius	
223	Temperature change with respect to a baseline ...	Degree Celsius	
224	Temperature change with respect to a baseline ...	Degree Celsius	

	Source	CTS_Code	\
0	Food and Agriculture Organization of the Unite...	ECCS	
1	Food and Agriculture Organization of the Unite...	ECCS	
2	Food and Agriculture Organization of the Unite...	ECCS	
3	Food and Agriculture Organization of the Unite...	ECCS	
4	Food and Agriculture Organization of the Unite...	ECCS	
..	
220	Food and Agriculture Organization of the Unite...	ECCS	
221	Food and Agriculture Organization of the	ECCS	Unite...

222 Food and Agriculture Organization of the Unite... ECCS
 223 Food and Agriculture Organization of the Unite... ECCS 224
 Food and Agriculture Organization of the Unite... ECCS

CTS_Name \
 0 Surface Temperature Change
 1 Surface Temperature Change
 2 Surface Temperature Change
 3 Surface Temperature Change
 4 Surface Temperature Change

 220 Surface Temperature Change
 221 Surface Temperature Change
 222 Surface Temperature Change
 223 Surface Temperature Change
 224 Surface Temperature Change

CTS_Full_Descriptor ... F2013 F2014 \
 0 Environment, Climate Change, Climate Indicator... ... 1.281 0.456
 1 Environment, Climate Change, Climate Indicator... ... 1.333 1.198
 2 Environment, Climate Change, Climate Indicator... ... 1.192 1.690
 3 Environment, Climate Change, Climate Indicator... ... 1.257 1.170
 4 Environment, Climate Change, Climate Indicator... ... 0.831 1.946

 220 Environment, Climate Change, Climate Indicator... ... 1.423 1.401
 221 Environment, Climate Change, Climate Indicator... ... 1.016 1.053 222
 Environment, Climate Change, Climate Indicator... ... NaN NaN
 223 Environment, Climate Change, Climate Indicator... ... 0.790 0.917
 224 Environment, Climate Change, Climate Indicator... ... 0.118 0.025

F2015 F2016 F2017 F2018 F2019 F2020 F2021 F2022
 0 1.093 1.555 1.540 1.544 0.910 0.498 1.327
 2.012 1 1.569 1.464 1.121 2.028 1.675 1.498 1.536 1.518 2
 1.121 1.757 1.512 1.210 1.115 1.926 2.330
 1.688 3 1.009 1.539 1.435 1.189 1.539 1.430 1.268
 1.256
 4 1.690 1.990 1.925 1.919 1.964 2.562 1.533 3.243

 220 1.510 1.732 2.204 0.942 1.477 2.069 1.593 1.970
 221 1.412 1.660 1.429 1.290 1.444 1.711 1.447 1.394 222 NaN
 NaN NaN NaN NaN NaN NaN NaN
 223 1.450 1.401 0.105 0.648 0.855 0.891 0.822 0.686

```
224    0.970 1.270 0.088 0.453 0.925 0.389 -0.125 -0.490
```

```
[225 rows x 32 columns]
```

3.0.2 About the Dataset

- This Climate Change dataset contains nearly 225 observations from across the World , with 32 columns of data.
- The Data includes both categorical and numeric values, providing a diverse range of information about the listings.
- This Dataset may be useful for analyzing trends and patterns of Climate Change across the World and also gain insights about different Climate Areas in various parts of the World.
- This dataset contains information about Climate Change Indicators from across the World from 2001 to 2022. By analyzing this data, you may be able to understand the trends and patterns of Climate Change across the World.

3.0.3 Understanding the given Column names :

- **ObjectId** : Unique Identifiers given to each Country
- **Country**: Name of the Country
- **ISO2** : Gives two-letter country code (ISO 3166-1 alpha-2) for each country.
- **ISO3** : Gives three-letter country code (ISO 3166-1 alpha-3) for each country.
- **Temp_Indicator** :Category of temperature change being measured related to a specific baseline of 15 degree celsius
- **Unit_of_Temp** : Unit of measurement for the temperature change in degree celsius
- **Source** :Indicates the organization responsible for collecting and providing the temperature change data.
- **CTS_Code** : Categorization system related to climate or temperature change.
- **CTS_Name**: Name with the climate or temperature change indicator
- **CTS_Full_Descriptor**: Contains a detailed description of the climate or temperature change indicator, providing additional info
- **Mean_Temp_Change** : Gives average temperature change for every Country across this Column
- **Max_Temp_Change** : Gives max temperature change for each Country
- **Min_Temp_Change** : Gives min temperature change for each Country
- **Total_Temp_Change** : Gives sum of temperature differences from baseline temperature of 15 degree celsius of all the years

4 Data Exploration and Data Cleaning

```
[2]: #Gives collection of all the columns
Climate_df.columns
```

```
[2]: Index(['ObjectId', 'Country', 'ISO2', 'ISO3', 'Indicator', 'Unit',
'Source',
          'CTS_Code', 'CTS_Name', 'CTS_Full_Descriptor', 'F2001', 'F2002',
```

```
'F2003', 'F2004', 'F2005', 'F2006', 'F2007', 'F2008', 'F2009',
'F2010',
'F2011', 'F2012', 'F2013', 'F2014', 'F2015', 'F2016', 'F2017',
'F2018',
'F2019', 'F2020', 'F2021', 'F2022'], dtype='object')
```

```
[3]: #Renames the given column names rename_col =
{'Indicator': 'Temp_Indicator', 'Unit': 'Unit_of_Temp'}
Climate_df = Climate_df.rename(columns = rename_col)
Climate_df.head(2)
```

```
[3]: ObjectId          Country ISO2 ISO3 \
0          1 Afghanistan, Islamic Rep. of      AF AFG
1          2 Albania AL ALB

          Temp_Indicator Unit_of_Temp \
0  Temperature change with respect to a baseline ... Degree Celsius
1  Temperature change with respect to a baseline ... Degree Celsius

          Source CTS_Code \
0 Food and Agriculture Organization of the Unite...
  ECCS 1 Food and Agriculture Organization of the
Unite...      ECCS

          CTS_Name \
0  Surface Temperature Change
1  Surface Temperature Change

          CTS_Full_Descriptor ... F2013 F2014 \
0  Environment, Climate Change, Climate Indicator... ... 1.281 0.456
1  Environment, Climate Change, Climate Indicator... ... 1.333 1.198

          F2015 F2016 F2017 F2018 F2019 F2020 F2021 F2022
0  1.093 1.555 1.540 1.544 0.910 0.498 1.327 2.012
1  1.569 1.464 1.121 2.028 1.675 1.498 1.536 1.518

[2 rows x 32 columns]
```

```
[4]: #Gives the shape (no. of rows,no. of columns) of dataset
Climate_df.shape
```

```
[4]: (225, 32)
```

```
[5]: Climate_df.info()
```

```
<class
'pandas.core.frame.DataFrame'>
```

RangeIndex: 225 entries, 0 to
 224 Data columns (total 32
 columns):

#	Column	Non-Null Count	Dtype
0	ObjectId	225 non-null	int64
1	Country	225 non-null	object
2	ISO2	223 non-null	object
3	ISO3	225 non-null	object
4	Temp_Indicator	225 non-null	object
5	Unit_of_Temp	225 non-null	object
6	Source	225 non-null	object
7	CTS_Code	225 non-null	object
8	CTS_Name	225 non-null	object
9	CTS_Full_Descriptor	225 non-null	object
10	F2001	208 non-null	float64
11	F2002	212 non-null	float64
12	F2003	214 non-null	float64
13	F2004	213 non-null	float64
14	F2005	212 non-null	float64
15	F2006	215 non-null	float64
16	F2007	217 non-null	float64
17	F2008	212 non-null	float64
18	F2009	212 non-null	float64
19	F2010	215 non-null	float64
20	F2011	217 non-null	float64
21	F2012	215 non-null	float64
22	F2013	216 non-null	float64
23	F2014	216 non-null	float64
24	F2015	216 non-null	float64
25	F2016	213 non-null	float64
26	F2017	214 non-null	float64
27	F2018	213 non-null	float64
28	F2019	213 non-null	float64
29	F2020	212 non-null	float64
30	F2021	213 non-null	float64
31	F2022	213 non-null	float64

dtypes: float64(22), int64(1),
 object(9) memory usage: 56.4+ KB

```
[6]: #check for any duplicate values and drop those duplicates
Climate_df = Climate_df.drop_duplicates()
Climate_df.count()
```

```
[6]: ObjectId      225
Country           225
ISO2              223
```

ISO3	225
Temp_Indicator	225
Unit_of_Temp	225
Source	225
CTS_Code	225
CTS_Name	225
CTS_Full_Descriptor	225
F2001	208
F2002	212
F2003	214
F2004	213
F2005	212
F2006	215
F2007	217
F2008	212
F2009	212
F2010	215
F2011	217
F2012	215
F2013	216
F2014	216
F2015	216
F2016	213
F2017	214
F2018	213
F2019	213
F2020	212
F2021	213
F2022	213 dtype:
	int64

```
[7]: #Check for any null or NaN values in the dataset and taking the sum
      of those_ null values
Climate_df.isnull().sum()
```

```
[7]: ObjectId      0
      Country      0
      ISO2         2
      ISO3         0
      Temp_Indicator 0
      Unit_of_Temp  0
      Source       0
      CTS_Code     0
```

CTS_Name	0
CTS_Full_Descriptor	0
F2001	17
F2002	13
F2003	11
F2004	12
F2005	13
F2006	10
F2007	8 F2008 13
F2009	13
F2010	10
F2011	8 F2012 10
F2013	9
F2014	9
F2015	9
F2016	12
F2017	11
F2018	12
F2019	12
F2020	13
F2021	12 F2022 12 dtype:

int64

```
[8]: #Replacing given numerical value columns null values with 0
      values columns_to_replace = ['F2001', 'F2002',
      'F2003', 'F2004', 'F2005', 'F2006', 'F2007', 'F2008', 'F2009', 'F2010',
      'F2011', 'F2012', 'F2013', 'F2014', 'F2015',
      'F2016', 'F2017', 'F2018', 'F2019', 'F2020',
      'F2021', 'F2022']
```

```
# Replace NaN values with 0 in the specified columns
```

```
Climate_df[columns_to_replace] = Climate_df[columns_to_replace].fillna(0)
```

```
[9]: #Replacing given categorical value columns null values with 'unknown'
      values
```

```
Climate_df['ISO2'].fillna('unknown', inplace=True)
```



```

Climate_df['ISO2'].isnull().sum()
[9]: 0
[10]: #Again checking for null values and this time there are no null values remaining Climate_df.isnull().sum()
[10]: ObjectId                0
Country                      0
ISO2                         0
ISO3                         0
Temp_Indicator               0
Unit_of_Temp                 0
Source                       0
CTS_Code                     0
CTS_Name                     0
CTS_Full_Descriptor          0
F2001                        0
F2002                        0
F2003                        0
F2004                        0
F2005                        0
F2006                        0
F2007                        0
F2008                        0
F2009                        0
F2010                        0
F2011                        0
F2012                        0
F2013                        0
F2014                        0
F2015                        0
F2016                        0
F2017                        0
F2018                        0
F2019                        0
F2020                        0
F2021                        0
F2022 dtype:                0
int64

[11]: #Gettting a random Sample of 5 elements from the dataset after Cleaning the dataset
Climate_df.sample(5)
[11]:   ObjectId      Country ISO2 ISO3 \

```

175 176 Seychelles SC SYC 14 15 Bahamas,
 The BS BHS
 2 3 Algeria DZ DZA
 101 102 Kazakhstan, Rep. of KZ KAZ
 219 220 West Bank and Gaza PS PSE

Temp_Indicator Unit_of_Temp \
 175 Temperature change with respect to a baseline ... Degree Celsius
 14 Temperature change with respect to a baseline ... Degree Celsius
 2 Temperature change with respect to a baseline ... Degree Celsius
 101 Temperature change with respect to a baseline ... Degree Celsius
 219 Temperature change with respect to a baseline ... Degree Celsius

Source CTS_Code \
 175 Food and Agriculture Organization of the ECCS
 Unite...
 14 Food and Agriculture Organization of the ECCS
 Unite...
 2 Food and Agriculture Organization of the Unite... ECCS
 101 Food and Agriculture Organization of the ECCS
 Unite...
 219 Food and Agriculture Organization of the ECCS
 Unite...

CTS_Name \
 175 Surface Temperature Change
 14 Surface Temperature Change
 2 Surface Temperature Change
 101 Surface Temperature Change
 219 Surface Temperature Change

CTS_Full_Descriptor ... F2013 F2014 \
 175 Environment, Climate Change, Climate Indicator... ... 0.749 0.863
 14 Environment, Climate Change, Climate Indicator... ... 0.565 0.883
 2 Environment, Climate Change, Climate Indicator... ... 1.192 1.690
 101 Environment, Climate Change, Climate Indicator... ... 1.621 0.673
 219 Environment, Climate Change, Climate Indicator... ... 1.114 0.863

F2015 F2016 F2017 F2018 F2019 F2020 F2021 F2022
 175 1.169 1.100 1.184 0.917 1.377 1.434 1.032 0.872
 14 1.114 1.042 1.331 1.023 1.443 1.611 0.879 1.480
 2 1.121 1.757 1.512 1.210 1.115 1.926 2.330 1.688
 101 1.609 2.240 1.757 0.641 1.487 2.853 1.465 2.712
 219 1.326 1.615 0.735 2.007 1.204 1.455 1.787 1.074

[5 rows x 32 columns]

```
[12]: #Gives summary statistics of Numerical columns of Climate Change
dataset Climate_df.describe()
```

```
[12]:      ObjectId  F2001      F2002      F2003      F2004      F2005 \
count 225.000000 225.000000 225.000000 225.000000 225.000000
      225.000000
mean   113.000000 0.785920      0.871556 0.802956 0.736356 0.803707
std     65.096083 0.505674      0.426329 0.458419 0.401072 0.411579
min       1.000000 -0.186000      0.000000 -0.252000 -0.622000 -0.393000
25%      57.000000 0.459000      0.646000 0.552000 0.505000 0.542000
50%     113.000000 0.689000      0.827000 0.822000 0.703000 0.827000
75%     169.000000 1.237000      1.123000 1.037000 0.957000 1.047000
max     225.000000 1.992000      2.255000 2.328000 2.150000 2.201000

      F2006      F2007      F2008      F2009 ...      F2013 \
count 225.000000 225.000000 225.000000 225.000000 ... 225.000000
mean    0.837618 0.986191 0.761702 0.857956 ...      0.893951
std     0.448922 0.569053 0.510197 0.428091 ...      0.364300
min     -0.505000 -0.219000 -0.139000 -0.319000 ...      0.000000
25%      0.581000 0.650000 0.406000 0.635000 ...      0.707000
50%      0.810000 0.903000 0.667000 0.869000 ...      0.885000
75%      1.109000 1.202000 1.090000 1.162000 ...      1.182000
max      2.343000 2.729000 2.607000 1.774000 ...      1.643000

      F2014      F2015      F2016      F2017      F2018      F2019 \
count 225.000000 225.000000 225.000000 225.000000 225.000000
      225.000000 mean 1.070222 1.218982 1.362747 1.218169 1.232667
      1.366098 std 0.595173 0.516915 0.507294 0.473529 0.650431 0.558986
min -0.092000 -0.430000 0.000000      0.000000 0.000000 0.000000
25%      0.704000 0.970000 1.097000      0.967000 0.823000 1.078000
50%      0.960000 1.201000 1.411000      1.257000 1.101000 1.396000
75%      1.306000 1.516000 1.692000      1.512000 1.609000 1.675000
```

```

max      2.704000  2.613000  2.459000      2.493000  2.772000  2.689000

      F2020      F2021      F2022
count      225.000000      225.000000
225.000000 mean  1.462364  1.271876
1.308400  std   0.704322  0.560254
0.721674  min   0.000000  -0.425000
1.305000  25%   1.128000  0.952000
0.834000
50%  1.430000  1.300000  1.268000  75%
1.778000  1.596000  1.865000 max
3.691000  2.676000  3.243000
[8 rows x 23 columns]

```

4.0.1 Adding Mean Temperature Change Column

```

[13]: mean_temp_change = Climate_df[['F2001', 'F2002', 'F2003', 'F2004',
'F2005',
    'F2006', 'F2007', 'F2008', 'F2009', 'F2010',
'F2011', 'F2012', 'F2013', 'F2014', 'F2015', 'F2016', 'F2017',
'F2018',
    'F2019', 'F2020', 'F2021', 'F2022']].mean(axis=1)

# Add the mean temperature change as a new column to the DataFrame
Climate_df['Mean_Temp_Change'] = mean_temp_change

# Display the updated DataFrame
Climate_df.head(2)

```

```

[13]: ObjectId      Country ISO2 ISO3 \
0      1 Afghanistan, Islamic Rep. of      AF AFG
1      2 Albania AL ALB

      Temp_Indicator Unit_of_Temp \
0  Temperature change with respect to a baseline ... Degree Celsius
1  Temperature change with respect to a baseline ... Degree Celsius

      Source CTS_Code \
0  Food and Agriculture Organization of the Unite...
  ECCS 1 Food and Agriculture Organization of the
  Unite...  ECCS

      CTS_Name \
0  Surface Temperature Change
1  Surface Temperature Change

      CTS_Full_Descriptor ... F2014 F2015 \

```

```

0 Environment, Climate Change, Climate Indicator... ... 0.456 1.093
1 Environment, Climate Change, Climate Indicator... ... 1.198 1.569

```

```

      F2016 F2017 F2018 F2019 F2020 F2021 F2022 Mean_Temp_Change
0  1.555  1.540  1.544  0.910  0.498  1.327  2.012  1.112727
1  1.464  1.121  2.028  1.675  1.498  1.536  1.518  1.174955

```

```
[2 rows x 33 columns]
```

```
[14]: #Calculating the value counts of Mean_Temp_Change
Climate_df['Mean_Temp_Change'].value_counts()
```

```

[14]: Mean_Temp_Change
      0.000000  5
0.749318  3
      1.054455  2
      0.988091  2
      1.112727  1
..
      1.540091  1
      0.776318  1
      0.773409  1
      0.831864  1
      0.301364  1
Name: count, Length: 217, dtype: int64

```

4.0.2 Max and Min of Mean Temperature Change

```

[15]: # Find the country with the highest mean temperature change
country_highest_change =
Climate_df.loc[Climate_df['Mean_Temp_Change'].idxmax()]

# Display the country with the highest mean temperature change
print("Country with the highest mean temperature change:")
print(country_highest_change[['Country', 'Mean_Temp_Change']])

```

```

Country with the highest mean temperature change:
Country          Estonia, Rep. of
Mean_Temp_Change          1.784
Name: 62, dtype: object

```

```
[16]: # Find the country with the lowest mean temperature change
country_lowest_change =
Climate_df.loc[Climate_df['Mean_Temp_Change'].idxmin()]

# Display the country with the lowest mean temperature change
print("Country with the lowest mean temperature change:")
print(country_lowest_change[['Country', 'Mean_Temp_Change']])
```

Country with the lowest mean temperature change:
Country Pitcairn Islands Mean_Temp_Change -
0.043727
Name: 159, dtype: object

4.0.3 Max and Min Temperature Countries of Year 2022

```
[17]: # Find the index of the maximum temperature change for the year 2022
max_temp_index_2022 = Climate_df['F2022'].idxmax()

# Get the country with the highest temperature change in 2022
country_max_temp_2022 = Climate_df.loc[max_temp_index_2022, 'Country']

print("Country with the highest temperature change in 2022:", _
country_max_temp_2022)
```

Country with the highest temperature change in 2022: Andorra,
Principality of

```
[18]: # Find the index of the minimum temperature change for the year 2022
min_temp_index_2022 = Climate_df['F2022'].idxmin()

# Get the country with the lowest temperature change in 2022
country_min_temp_2022 = Climate_df.loc[min_temp_index_2022, 'Country']

print("Country with the lowest temperature change in 2022:", _
country_min_temp_2022)
```

Country with the lowest temperature change in 2022: Botswana

4.0.4 Adding Max, Min and Total Temperature Change Columns

```
[19]: # Find the maximum and minimum temperature change across all years for
each
country max_temp_change = Climate_df.iloc[:,
10:32].max(axis=1) min_temp_change =
Climate_df.iloc[:, 10:32].min(axis=1)

# Calculate the total temperature change for each country across all
years total_temp_change = Climate_df.iloc[:, 10:32].sum(axis=1)
```

```

# Add the results as new columns to the DataFrame
Climate_df['Max_Temp_Change'] = max_temp_change
Climate_df['Min_Temp_Change'] = min_temp_change
Climate_df['Total_Temp_Change'] = total_temp_change

# Display the DataFrame with the new columns

```

```

[19]:      ObjectId      Country  ISO2 ISO3 \
0          1 Afghanistan, Islamic Rep. of AF AFG
1          2      Albania AL ALB
2          3      Algeria DZ DZA
3          4  American Samoa AS ASM
4          5  Andorra, Principality of AD AND
..      ...
220 221 Western Sahara EH ESH 221 222 World unknown WLD
222 223 Yemen, Rep. of YE YEM
223      224      Zambia      ZM ZMB
224      225      Zimbabwe ZW ZWE

      Temp_Indicator Unit_of_Temp \
0  Temperature change with respect to a baseline ... Degree Celsius
1  Temperature change with respect to a baseline ... Degree Celsius
2  Temperature change with respect to a baseline ... Degree Celsius
3  Temperature change with respect to a baseline ... Degree Celsius
4  Temperature change with respect to a baseline ... Degree Celsius
..
220 Temperature change with respect to a baseline ... Degree Celsius
221 Temperature change with respect to a baseline ... Degree Celsius
222 Temperature change with respect to a baseline ... Degree Celsius
223 Temperature change with respect to a baseline ... Degree Celsius
224 Temperature change with respect to a baseline ... Degree Celsius

      Source CTS_Code \
0 Food and Agriculture Organization of the Unite... ECCS
1 Food and Agriculture Organization of the Unite... ECCS
2 Food and Agriculture Organization of the Unite... ECCS
3 Food and Agriculture Organization of the Unite... ECCS
4 Food and Agriculture Organization of the Unite... ECCS
..
220 Food and Agriculture Organization of the      ECCS
Unite...

```

221 Food and Agriculture Organization of the Unite...	ECCS
222 Food and Agriculture Organization of the Unite...	ECCS
223 Food and Agriculture Organization of the Unite...	ECCS
224 Food and Agriculture Organization of the Unite...	ECCS

	CTS_Name \
0	Surface Temperature Change
1	Surface Temperature Change
2	Surface Temperature Change
3	Surface Temperature Change
4	Surface Temperature Change
..	...
220	Surface Temperature Change
221	Surface Temperature Change
222	Surface Temperature Change
223	Surface Temperature Change
224	Surface Temperature Change

	CTS_Full_Descriptor ... F2017 F2018 \
0	Environment, Climate Change, Climate Indicator... ... 1.540 1.544
1	Environment, Climate Change, Climate Indicator... ... 1.121 2.028
2	Environment, Climate Change, Climate Indicator... ... 1.512 1.210
3	Environment, Climate Change, Climate Indicator... ... 1.435 1.189
4	Environment, Climate Change, Climate Indicator... ... 1.925 1.919
..
220	Environment, Climate Change, Climate Indicator... ... 2.204 0.942
221	Environment, Climate Change, Climate Indicator... ... 1.429 1.290
222	Environment, Climate Change, Climate Indicator... ... 0.000 0.000
223	Environment, Climate Change, Climate Indicator... ... 0.105 0.648
224	Environment, Climate Change, Climate Indicator... ... 0.088 0.453
	F2019 F2020 F2021 F2022 Mean_Temp_Change Max_Temp_Change \ 0 0.910
	0.498 1.327 2.012 1.112727 2.012
1	1.675 1.498 1.536 1.518 1.174955 2.028
2	1.115 1.926 2.330 1.688 1.456682 2.330
3	1.539 1.430 1.268 1.256 1.002273 1.539
4	1.964 2.562 1.533 3.243 1.542500 3.243
..


```

220 1.477 2.069 1.593 1.970      1.474182      2.204
221 1.444 1.711 1.447 1.394      1.177773      1.711
222 0.000 0.000 0.000 0.000      0.000000      0.000
223 0.855 0.891 0.822 0.686      0.784045      1.450
224 0.925 0.389 -0.125 -0.490    0.301364      1.270

```

```

      Min_Temp_Change
      Total_Temp_Change
0      0.223 24.480 1 0.189 25.849
2      0.945 32.047 3 0.000 22.050
4              0.471              33.935
..              ...              ...
220    0.903 32.432
221    0.834 25.911 222 0.000 0.000
223              0.105 17.249
224              -0.490 6.630

```

```
[225 rows x 36 columns]
```

4.0.5 Line Chart for Mean Temperature Change

```

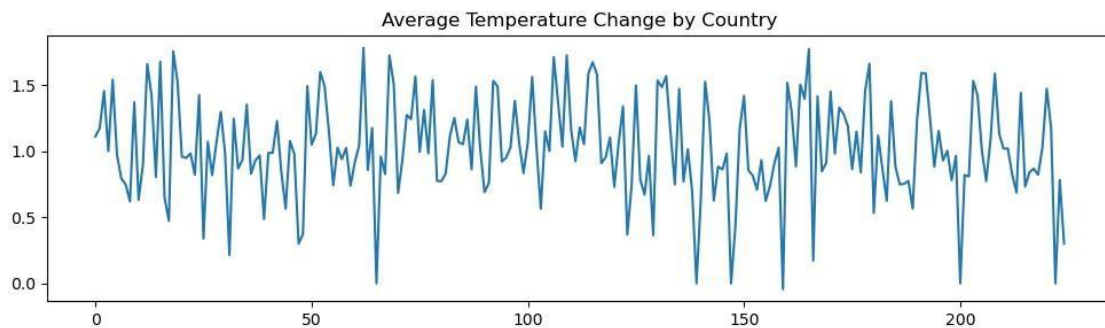
[20]: Climate_df['Mean_Temp_Change'].plot(kind='line',x='Country',
      figsize=(12, 3),
      title='Average Temperature Change by Country')

```

```

[20]: <Axes: title={'center': 'Average Temperature Change by Country'}>

```



4.0.6 Bar Chart for Top 10 Largest Mean Temperature by Countries

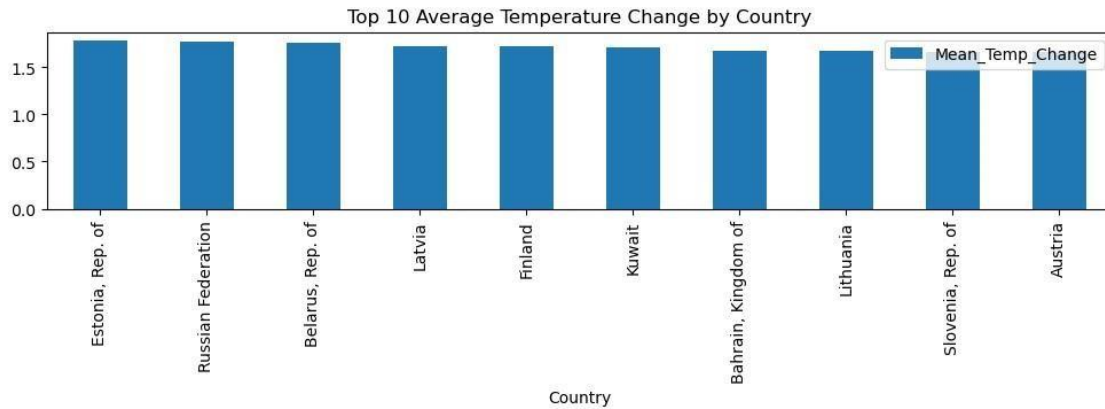
```

[26]: top_10_mean_temp_change = Climate_df.nlargest(10, 'Mean_Temp_Change')
      top_10_mean_temp_change.plot(kind='bar', x='Country',
      y='Mean_Temp_Change',

```

```
figsize=(12, 2), title='Top 10 Average Temperature Change by Country')
```

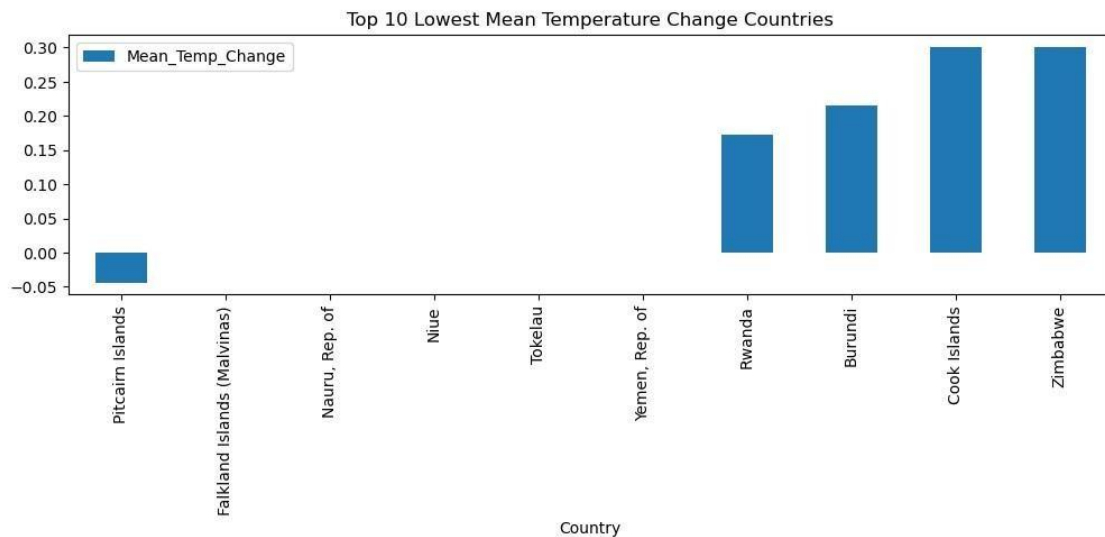
```
[26]: <Axes: title={'center': 'Top 10 Average Temperature Change by Country'}, xlabel='Country'>
```



4.0.7 'Bar Chart for Top 10 Lowest Mean Temperature Change Countries

```
[27]: top_10_min_mean_temp_change = Climate_df.nsmallest(10,
'Mean_Temp_Change') top_10_min_mean_temp_change.plot(kind='bar',
x='Country', y='Mean_Temp_Change',
figsize=(12, 3), title='Top 10 Lowest Mean Temperature Change Countries')
```

```
[27]: <Axes: title={'center': 'Top 10 Lowest Mean Temperature Change Countries'}, xlabel='Country'>
```



4.0.8 Bar Chart for Maximum Temperature Change

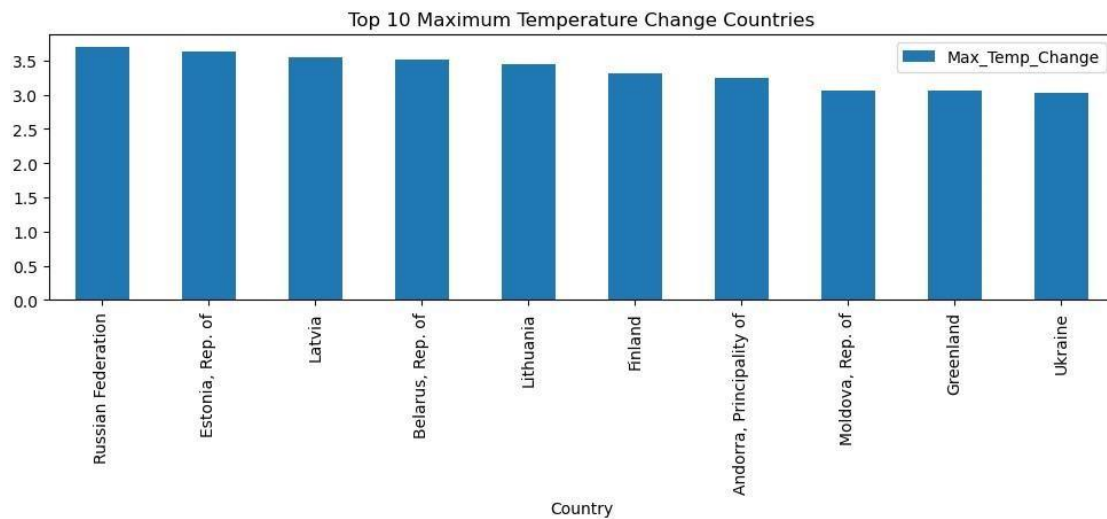
```
[28]: top_10_max_temp_change =
```

```
Climate_df.sort_values(by='Max_Temp_Change',
```

```
    ascending=False).head(10)
top_10_max_temp_change.plot(kind='bar', x='Country',
y='Max_Temp_Change',
```

```
    figsize=(12,3), title='Top 10 Maximum Temperature Change Countries')
```

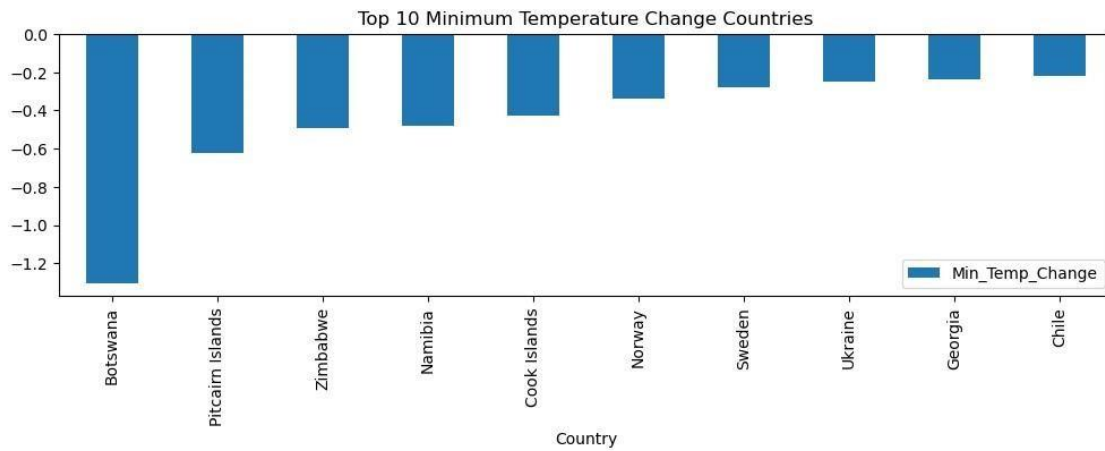
```
[28]: <Axes: title={'center': 'Top 10 Maximum Temperature Change
Countries'}, xlabel='Country'>
```



4.0.9 Bar Chart for Minimum Temperature Change

```
[29]: top_10_min_temp_change =
Climate_df.sort_values(by='Min_Temp_Change').head(10)
top_10_min_temp_change.plot(kind='bar', x='Country',
y='Min_Temp_Change',
    figsize=(12,3), title='Top 10 Minimum Temperature Change Countries')
```

```
[29]: <Axes: title={'center': 'Top 10 Minimum Temperature Change
Countries'}, xlabel='Country'>
```



4.1 Pivot Table :

```
[25]: #Making a Pivot Table with Column as Country and giving values of 4
      Columns of_ Numerical datatype
Climate_df.pivot_table(index='Country',values=[_
      'Mean_Temp_Change','Total_Temp_Change','Max_Temp_Change','Min_Temp_C
      hange'])
```

```
[25]:
```

	Max_Temp_Change	Mean_Temp_Change \
Country		
Afghanistan, Islamic Rep. of	2.012	1.112727
Albania	2.028	1.174955
Algeria	2.330	1.456682
American Samoa	1.539	1.002273
Andorra, Principality of	3.243	1.542500
...
Western Sahara	2.204	1.474182
World	1.711	1.177773
Yemen, Rep. of	0.000	0.000000
Zambia	1.450	0.784045
Zimbabwe	1.270	0.301364

	Min_Temp_Change	Total_Temp_Change
Country		
Afghanistan, Islamic Rep. of	0.223	24.480
Albania	0.189	25.849
Algeria	0.945	32.047
American Samoa	0.000	22.050
Andorra, Principality of	0.471	33.935
...

Western Sahara	0.903	32.432
World	0.834	25.911
Yemen, Rep. of	0.000	0.000
Zambia	0.105	17.249
Zimbabwe	-0.490	6.630

[225 rows x 4 columns]

5 Observations :

- The dataset consists of 225 rows, each representing a different country or region.
- The temperature changes are measured in degrees Celsius with respect to a baseline of 15 degree celsius.
- Mean_Temp_Change, Max_Temp_Change, Min_Temp_Change, and Total_Temp_Change columns seem to provide summary statistics for temperature changes.
- We found the the country with the highest mean temperature change is **Estonia** with **1.784 degree celsius** rise from baseline. We found the country with the lowest mean temperature change is **Pitcairn Islands** with **-0.043727 degree celsius** fall from baseline.
- The **Maximum temperature change** for the year **2022** is of Country **Andorra**.
- The **Minimum temperature change** for the year **2022** is of Country **Botswana**.
- The Top 10 Average Maximum Temperature by Country which were most affected were: **Estonia,Russia,Belarus,Latvia,Finland,Kuwait,Bahrain,Lithuania,Slovenia,Austria** •
- The Top 10 Average Minimum Temperature by Country which were least affected were: **Pitcairn Islands,Falkland,Nauru,Niue,Yemen,Rwanda,Burundi,Cook Islands,Zimbabwe**
- Top 10 Countries where max Temperature change across past 22 Years: **Estonia,Russia,Belarus,Latvia,Finland,Andorra,Moldova,Lithuania,Greenland,Ukraine**
- Top 10 Countries where min Temperature change across past 22 Years: **Pitcairn Islands,Botswana,Namibia,Norway,Sweden,Ukraine,Georgia,Chile,Cook Islands**
- From the above Max Temperature Change and Max Mean Temperature Change values we can see that:Countries further away from the Equator are getting the most affected by the rise in Temperature.
 - Also,From the above Min Temperature Change and Min Mean Tempoeature Change values we can see that:Countries closer to the Equator are getting the least affected by the rise in Temperature.

6 Conclusion :

- There is a significant variation in temperature changes across different countries and regions. Some areas have experienced considerable increases in temperature, while others have seen more moderate changes or even decreases.
- It's notable that the mean temperature change across all countries is positive, indicating a general trend of warming temperatures globally.
- The maximum and minimum temperature changes provide insights into the range of variation within each country or region.

- Further analysis could involve investigating correlations between temperature changes and other factors such as geographical location, population density, and economic activities to better understand the drivers of climate change impacts.
- The data highlights the urgent need for global action to mitigate climate change and its associated impacts, as rising temperatures can have significant consequences for ecosystems, agriculture, water resources, and human well-being.

7 The End