

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
df=pd.read_csv("C:/Users/chalanagowda/Downloads/climate_change_impact_on_agricultur
print(df)
```

	Year	Country	Region	Crop_Type	Average_Temperature_C	\
0	2001	India	West Bengal	Corn	1.55	
1	2024	China	North	Corn	3.23	
2	2001	France	Ile-de-France	Wheat	21.11	
3	2001	Canada	Prairies	Coffee	27.85	
4	1998	India	Tamil Nadu	Sugarcane	2.19	
...	
9995	2022	France	Nouvelle-Aquitaine	Cotton	30.48	
9996	1999	Australia	Queensland	Soybeans	9.53	
9997	2000	Argentina	Patagonia	Coffee	31.92	
9998	1996	Brazil	Southeast	Soybeans	13.95	
9999	2015	China	South	Corn	11.78	

	Total_Precipitation_mm	CO2_Emissions_MT	Crop_Yield_MT_per_HA	\
0	447.06	15.22	1.737	
1	2913.57	29.82	1.737	
2	1301.74	25.75	1.719	
3	1154.36	13.91	3.890	
4	1627.48	11.81	1.080	
...	
9995	685.93	17.64	3.033	
9996	2560.38	10.68	2.560	
9997	357.76	26.01	1.161	
9998	1549.52	17.31	3.348	
9999	1676.25	5.34	3.710	

	Extreme_Weather_Events	Irrigation_Access_%	Pesticide_Use_KG_per_HA	\
0	8	14.54	10.08	
1	8	11.05	33.06	
2	5	84.42	27.41	
3	5	94.06	14.38	
4	9	95.75	44.35	
...	
9995	9	27.56	41.96	
9996	4	77.02	5.45	
9997	10	78.53	11.94	
9998	2	42.65	44.71	
9999	5	46.41	48.28	

	Fertilizer_Use_KG_per_HA	Soil_Health_Index	Adaptation_Strategies	\
0	14.78	83.25	Water Management	
1	23.25	54.02	Crop Rotation	
2	65.53	67.78	Water Management	
3	87.58	91.39	No Adaptation	
4	88.08	49.61	Crop Rotation	
...	
9995	10.95	43.41	No Adaptation	
9996	82.32	59.39	No Adaptation	
9997	26.00	41.46	Water Management	
9998	25.07	75.10	Crop Rotation	
9999	98.27	59.38	Water Management	

	Economic_Impact_Million_USD
0	808.13
1	616.22
2	796.96

```
3          790.32
4          401.72
...          ...
9995       1483.06
9996        829.61
9997        155.99
9998       1613.90
9999        453.14
```

```
[10000 rows x 15 columns]
```

```
In [ ]:
```

```
In [2]: #Information about the dataset
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 15 columns):
 #   Column                                  Non-Null Count  Dtype
---  -
 0   Year                                  10000 non-null  int64
 1   Country                             10000 non-null  object
 2   Region                              10000 non-null  object
 3   Crop_Type                           10000 non-null  object
 4   Average_Temperature_C               10000 non-null  float64
 5   Total_Precipitation_mm              10000 non-null  float64
 6   CO2_Emissions_MT                    10000 non-null  float64
 7   Crop_Yield_MT_per_HA                10000 non-null  float64
 8   Extreme_Weather_Events              10000 non-null  int64
 9   Irrigation_Access_%                 10000 non-null  float64
10   Pesticide_Use_KG_per_HA             10000 non-null  float64
11   Fertilizer_Use_KG_per_HA            10000 non-null  float64
12   Soil_Health_Index                   10000 non-null  float64
13   Adaptation_Strategies               10000 non-null  object
14   Economic_Impact_Million_USD         10000 non-null  float64
dtypes: float64(9), int64(2), object(4)
memory usage: 1.1+ MB
```

```
In [3]: df.duplicated()
```

```
Out[3]: 0      False
        1      False
        2      False
        3      False
        4      False
        ...
        9995   False
        9996   False
        9997   False
        9998   False
        9999   False
Length: 10000, dtype: bool
```

```
In [6]: df.duplicated().sum()
```

Out[6]: 0

```
In [8]: #Display first 5 rows
df.head()
```

Out[8]:

	Year	Country	Region	Crop_Type	Average_Temperature_C	Total_Precipitation_mm	CO2
0	2001	India	West Bengal	Corn	1.55	447.06	
1	2024	China	North	Corn	3.23	2913.57	
2	2001	France	Ile-de-France	Wheat	21.11	1301.74	
3	2001	Canada	Prairies	Coffee	27.85	1154.36	
4	1998	India	Tamil Nadu	Sugarcane	2.19	1627.48	

```
In [10]: #Display last 5 rows
df.tail()
```

Out[10]:

	Year	Country	Region	Crop_Type	Average_Temperature_C	Total_Precipitation_mm	CO2
9995	2022	France	Nouvelle-Aquitaine	Cotton	30.48	681.5	
9996	1999	Australia	Queensland	Soybeans	9.53	2561.5	
9997	2000	Argentina	Patagonia	Coffee	31.92	351.5	
9998	1996	Brazil	Southeast	Soybeans	13.95	1541.5	
9999	2015	China	South	Corn	11.78	1671.5	

```
In [11]: #The number of rows and columns
df.shape
```

Out[11]: (10000, 15)

```
In [12]: #checking for NULL values (Boolean values)
df.isnull()
```

Out[12]:

	Year	Country	Region	Crop_Type	Average_Temperature_C	Total_Precipitation_mm
0	False	False	False	False	False	False
1	False	False	False	False	False	False
2	False	False	False	False	False	False
3	False	False	False	False	False	False
4	False	False	False	False	False	False
...
9995	False	False	False	False	False	False
9996	False	False	False	False	False	False
9997	False	False	False	False	False	False
9998	False	False	False	False	False	False
9999	False	False	False	False	False	False

10000 rows × 7 columns

In [5]:

df.isnull().sum()

Out[5]:

Year	0
Country	0
Region	0
Crop_Type	0
Average_Temperature_C	0
Total_Precipitation_mm	0
CO2_Emissions_MT	0
Crop_Yield_MT_per_HA	0
Extreme_Weather_Events	0
Irrigation_Access_%	0
Pesticide_Use_KG_per_HA	0
Fertilizer_Use_KG_per_HA	0
Soil_Health_Index	0
Adaptation_Strategies	0
Economic_Impact_Million_USD	0
dtype:	int64

In [9]:

#data cleaning
#handling missing values
df.fillna(df.select_dtypes(include=[np.number]).mean(),inplace=True)

In [10]:

print(df.describe())

	Year	Average_Temperature_C	Total_Precipitation_mm	\
count	10000.000000	10000.000000	10000.000000	
mean	2007.088700	15.241299	1611.663834	
std	10.084245	11.466955	805.016815	
min	1990.000000	-4.990000	200.150000	
25%	1999.000000	5.430000	925.697500	
50%	2007.000000	15.175000	1611.160000	
75%	2016.000000	25.340000	2306.997500	
max	2024.000000	35.000000	2999.670000	

	CO2_Emissions_MT	Crop_Yield_MT_per_HA	Extreme_Weather_Events	\
count	10000.000000	10000.000000	10000.000000	
mean	15.246608	2.240017	4.980900	
std	8.589423	0.998342	3.165808	
min	0.500000	0.450000	0.000000	
25%	7.760000	1.449000	2.000000	
50%	15.200000	2.170000	5.000000	
75%	22.820000	2.930000	8.000000	
max	30.000000	5.000000	10.000000	

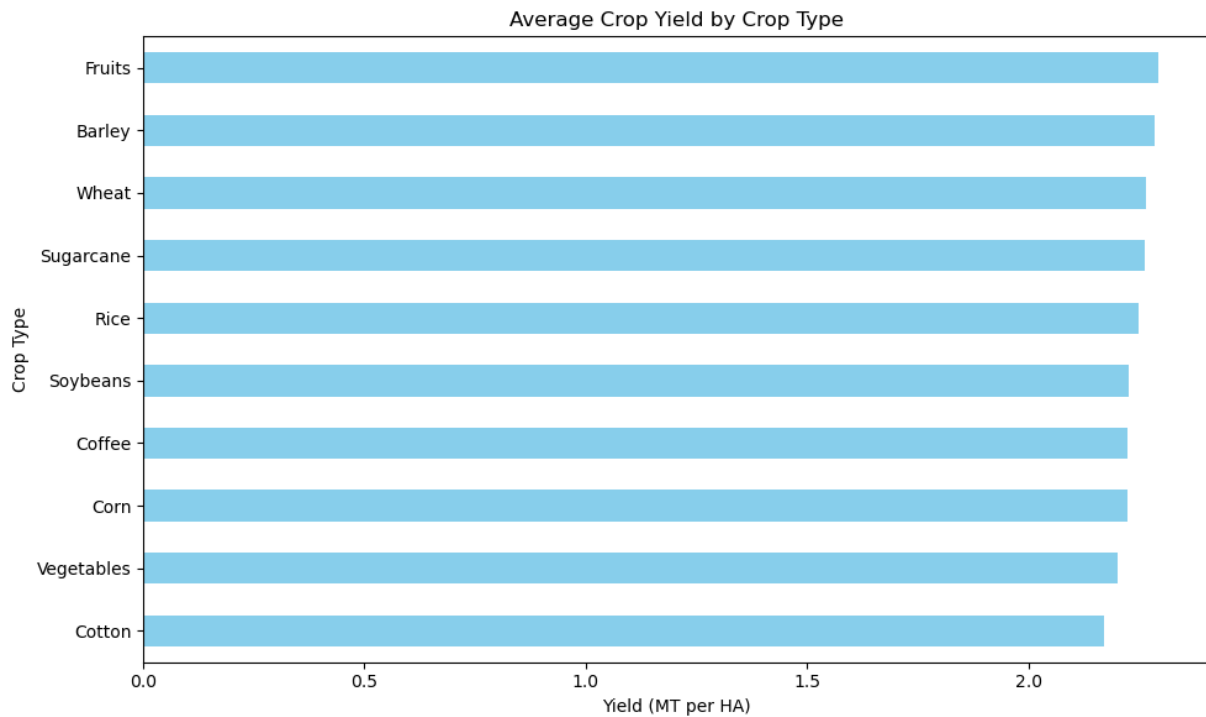
	Irrigation_Access_%	Pesticide_Use_KG_per_HA	Fertilizer_Use_KG_per_HA	\
count	10000.000000	10000.000000	10000.000000	
mean	55.248332	24.955735	49.973708	
std	25.988305	14.490962	28.711027	
min	10.010000	0.000000	0.010000	
25%	32.677500	12.527500	25.390000	
50%	55.175000	24.930000	49.635000	
75%	77.582500	37.470000	74.825000	
max	99.990000	49.990000	99.990000	

	Soil_Health_Index	Economic_Impact_Million_USD
count	10000.000000	10000.000000
mean	64.901278	674.269658
std	20.195882	414.591431
min	30.000000	47.840000
25%	47.235000	350.545000
50%	64.650000	583.920000
75%	82.472500	917.505000
max	100.000000	2346.470000

```
In [11]: #removing duplicates
df.drop_duplicates(inplace=True)
```

```
In [12]: #Bar Plot: Average Yield by Crop Type
avg_yield_crop = df.groupby('Crop_Type')['Crop_Yield_MT_per_HA'].mean().sort_values

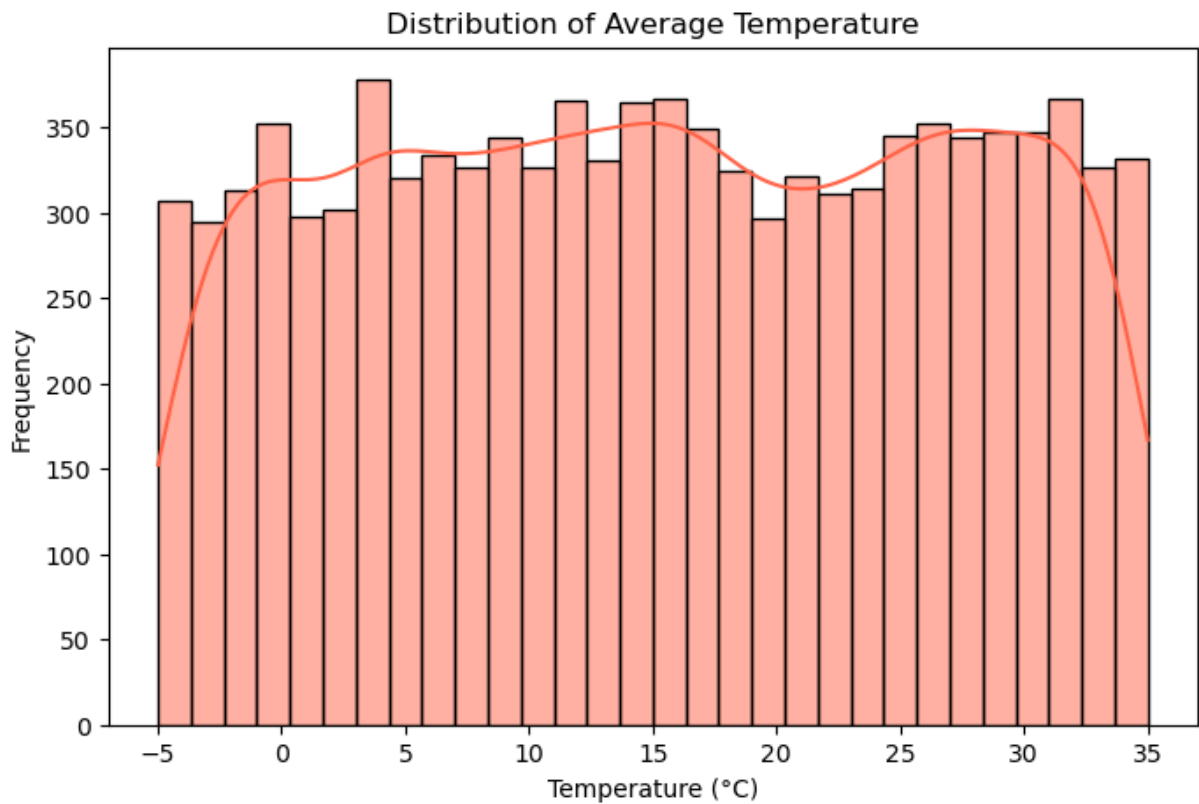
plt.figure(figsize=(10, 6))
avg_yield_crop.plot(kind='barh', color='skyblue')
plt.title('Average Crop Yield by Crop Type')
plt.xlabel('Yield (MT per HA)')
plt.ylabel('Crop Type')
plt.tight_layout()
plt.show()
```



```
In [21]: #Histogram: Distribution of Average Temperature
plt.figure(figsize=(8, 5))
sns.histplot(df['Average_Temperature_C'], bins=30, kde=True, color='tomato')
plt.title('Distribution of Average Temperature')
plt.xlabel('Temperature (°C)')
plt.ylabel('Frequency')
plt.show()
```

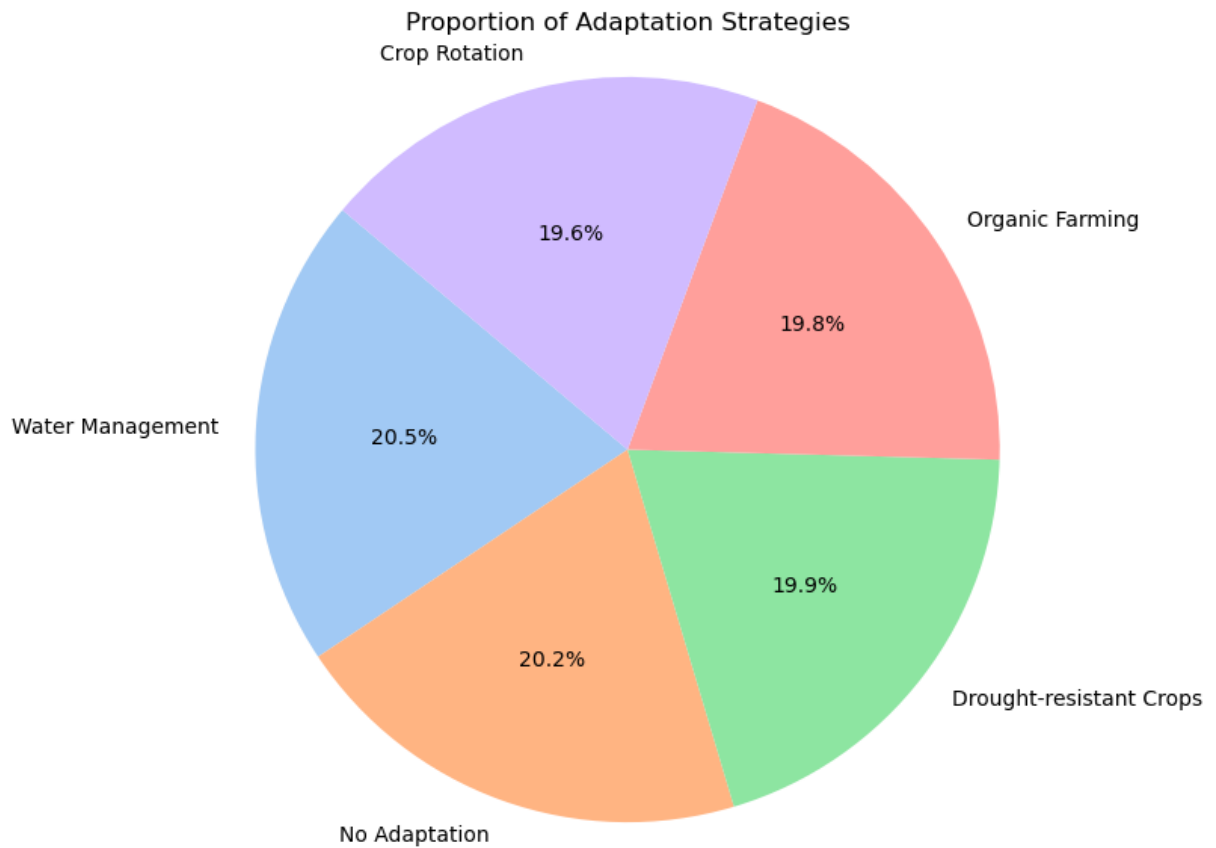
C:\Users\chalanagowda\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

```
with pd.option_context('mode.use_inf_as_na', True):
```



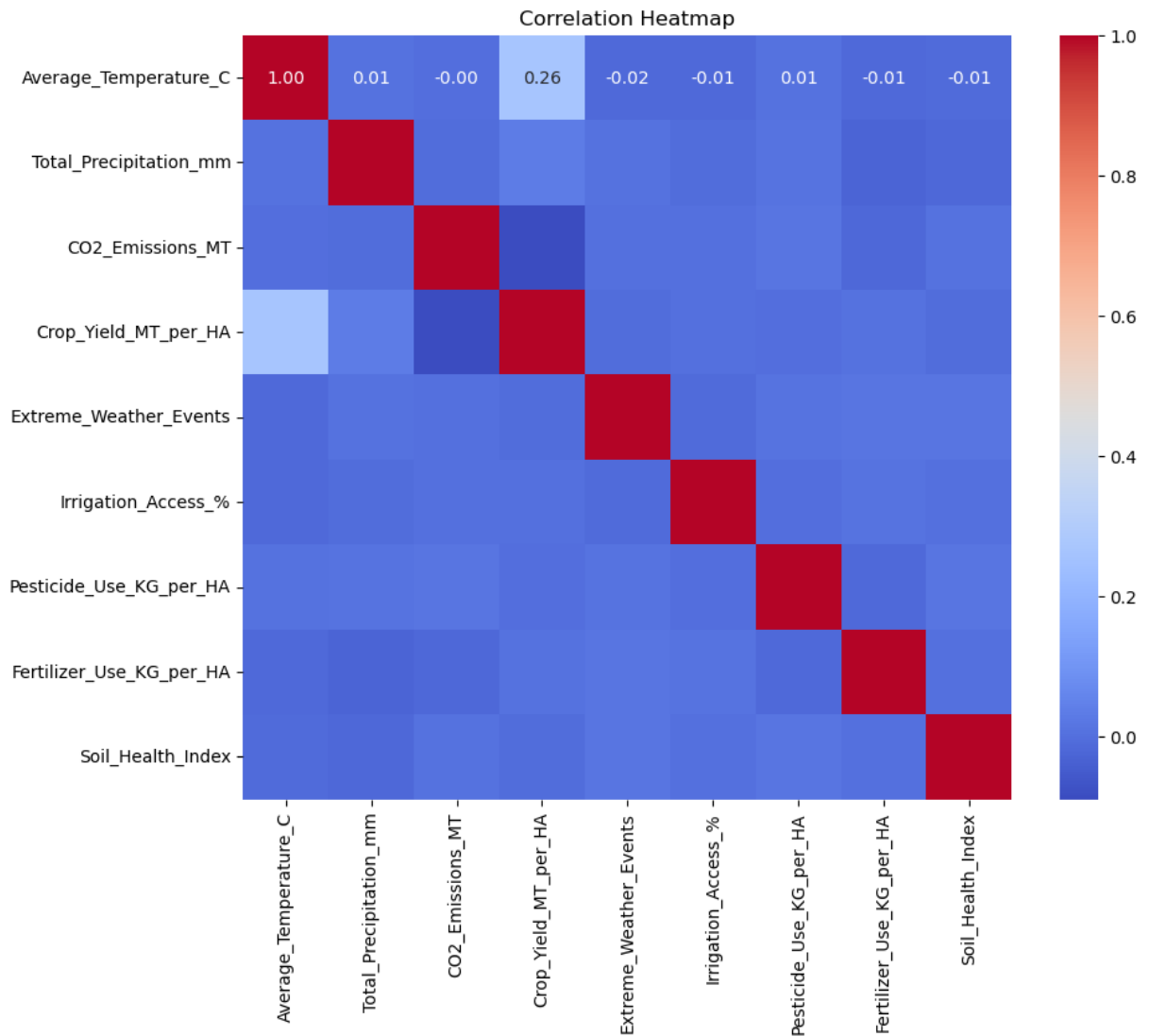
```
In [14]: #Pie Chart: Adaptation Strategies Proportion
strategy_counts = df['Adaptation_Strategies'].value_counts()

plt.figure(figsize=(7, 7))
plt.pie(strategy_counts, labels=strategy_counts.index, autopct='%1.1f%%', startangle=0)
plt.title('Proportion of Adaptation Strategies')
plt.axis('equal')
plt.show()
```

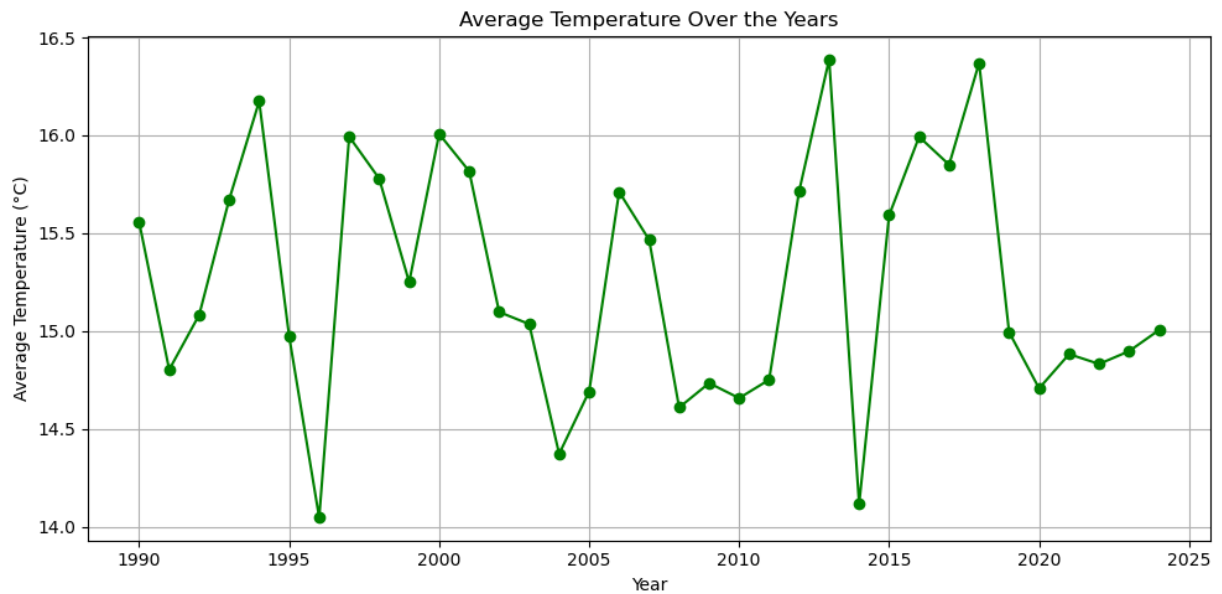
```
In [15]: #Correlation Heatmap (Climate vs Agriculture)
# Select numeric columns for correlation
corr_columns = ['Average_Temperature_C', 'Total_Precipitation_mm', 'CO2_Emissions_Mt',
                'Crop_Yield_MT_per_HA', 'Extreme_Weather_Events', 'Irrigation_Accessibility',
                'Pesticide_Use_KG_per_HA', 'Fertilizer_Use_KG_per_HA', 'Soil_Health_Index']

plt.figure(figsize=(10, 8))
sns.heatmap(df[corr_columns].corr(), annot=True, cmap='coolwarm', fmt=".2f")
plt.title('Correlation Heatmap')
plt.show()
```

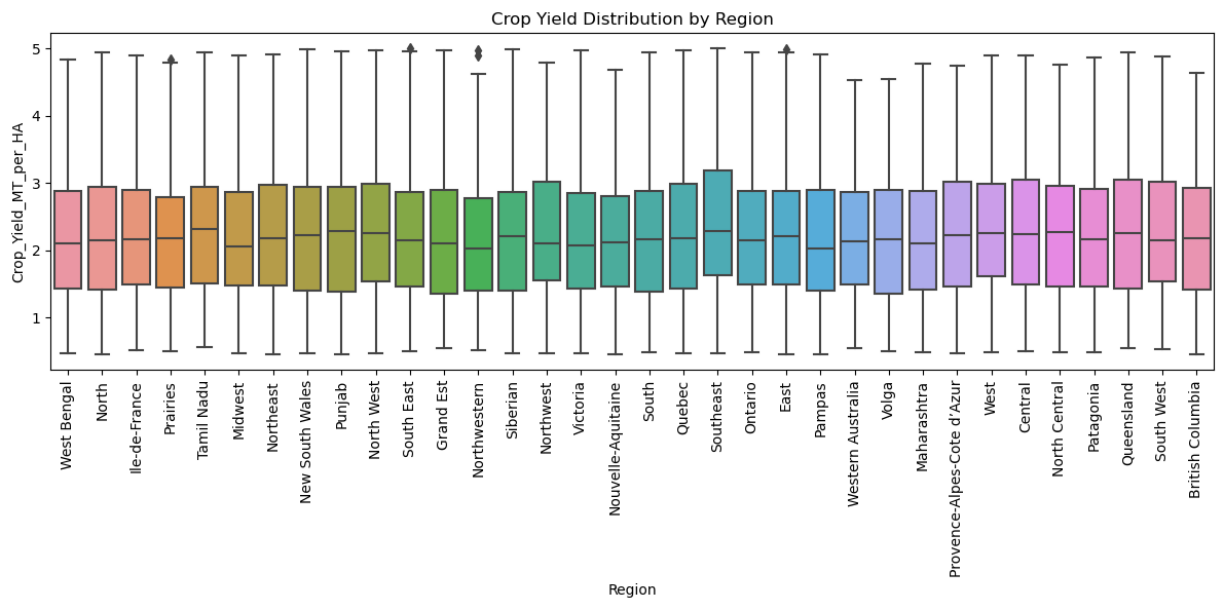


```
In [16]: #Time Series Plot: Average Temperature over Years
avg_temp_by_year = df.groupby('Year')['Average_Temperature_C'].mean()

plt.figure(figsize=(10, 5))
avg_temp_by_year.plot(marker='o', linestyle='--', color='green')
plt.title('Average Temperature Over the Years')
plt.xlabel('Year')
plt.ylabel('Average Temperature (°C)')
plt.grid(True)
plt.tight_layout()
plt.show()
```



```
In [22]: #Boxplot: Yield Distribution by Region
plt.figure(figsize=(12, 6))
sns.boxplot(data=df, x='Region', y='Crop_Yield_MT_per_HA')
plt.title('Crop Yield Distribution by Region')
plt.xticks(rotation=90)
plt.tight_layout()
plt.show()
```



```
In [ ]:
```