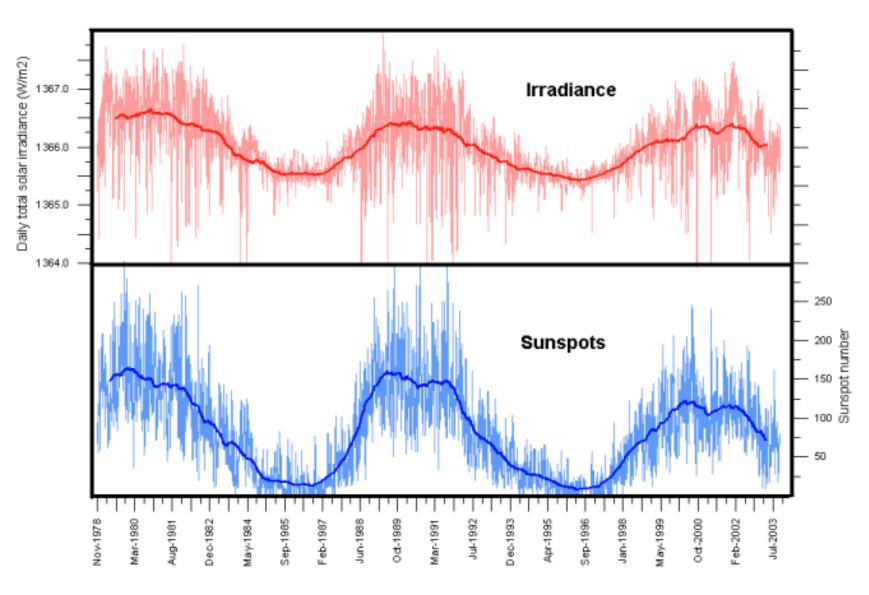


SUN: NOT STATIC.





Source: NASA



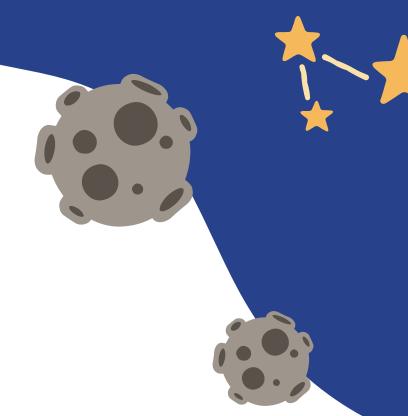




RESEARCH QUESTIONS

DEFINING THE OBJECTIVES AND GOALS OF THE PROJECT

Is there is any sort of relation between Earth's surface temperature and the Solar activity level? Are there any other factors?



DATASETS



01 Sunspot numbers

02 Global Temperature Global
Carbon
Emissions







SOLAR ACTIVITES EDA

TIME RANGE: MARCH 1749 - MARCH 2021

Monthly	v Mean	Total	Sunspot	Number
I TO THE THE	y inicali	IVIAI	Juliapol	Humber

	monany moan roun cameporname.
Date	
1749-01-31	96.7
1749-02-28	104.3
1749-03-31	116.7
1749-04-30	92.8
1749-05-31	141.7
2020-09-30	0.6
2020-10-31	14.4
2020-11-30	34.0
2020-12-31	21.8
2021-01-31	10.4

3265 rows × 1 columns

In [71]: print(sunspot_data.index.is_monotonic_increasing)
 sunspot_data.isnull().sum()

True

Out[71]: Monthly Mean Total Sunspot Number 0

dtype: int64



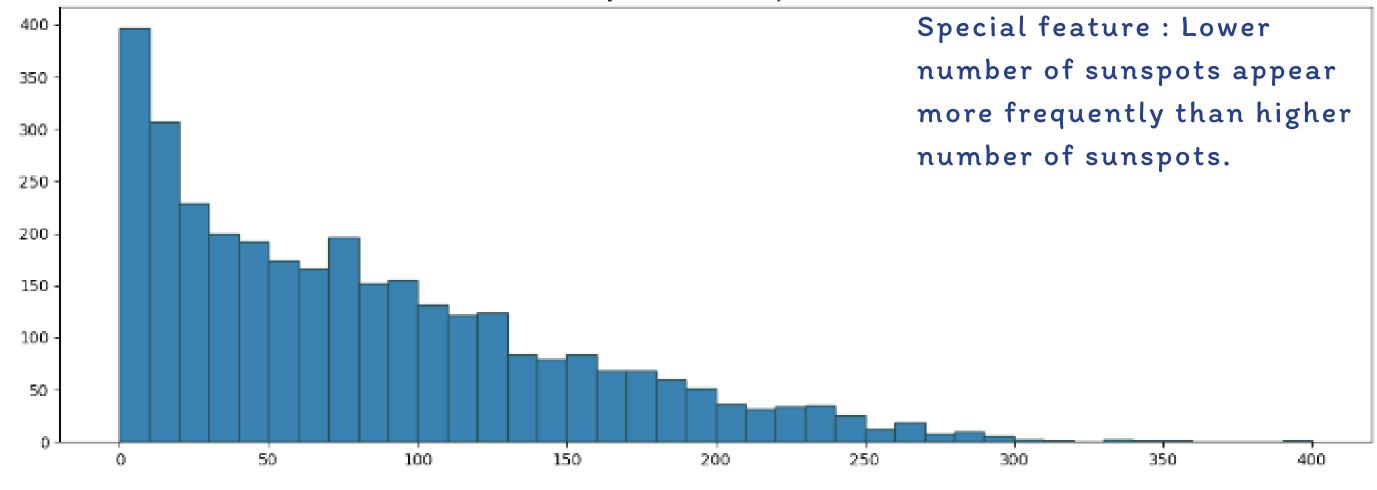


FREQUENCY DISTRIBUTION





Monthly Mean Total Sunspot Number

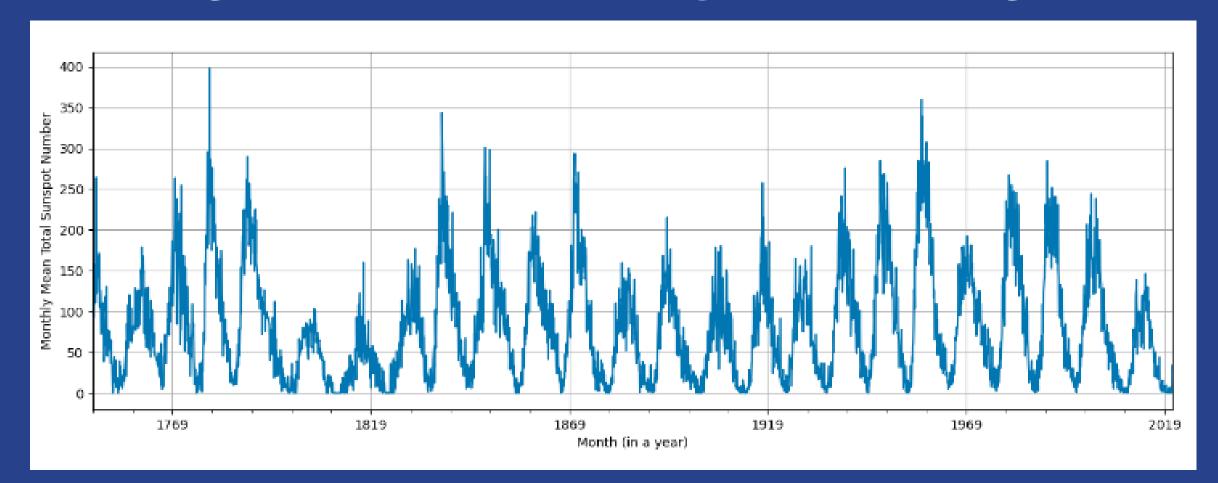


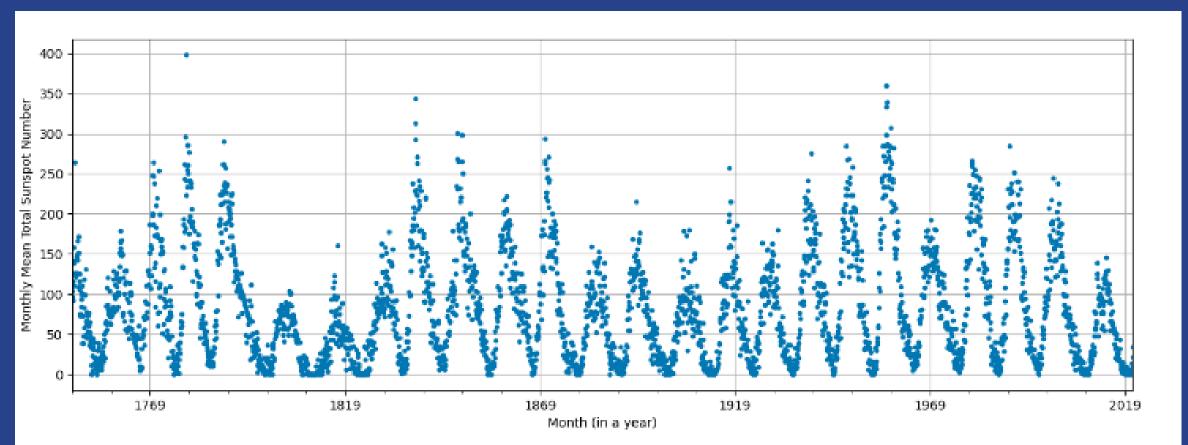




LINE GRAPH AND SCATTER GRAPH

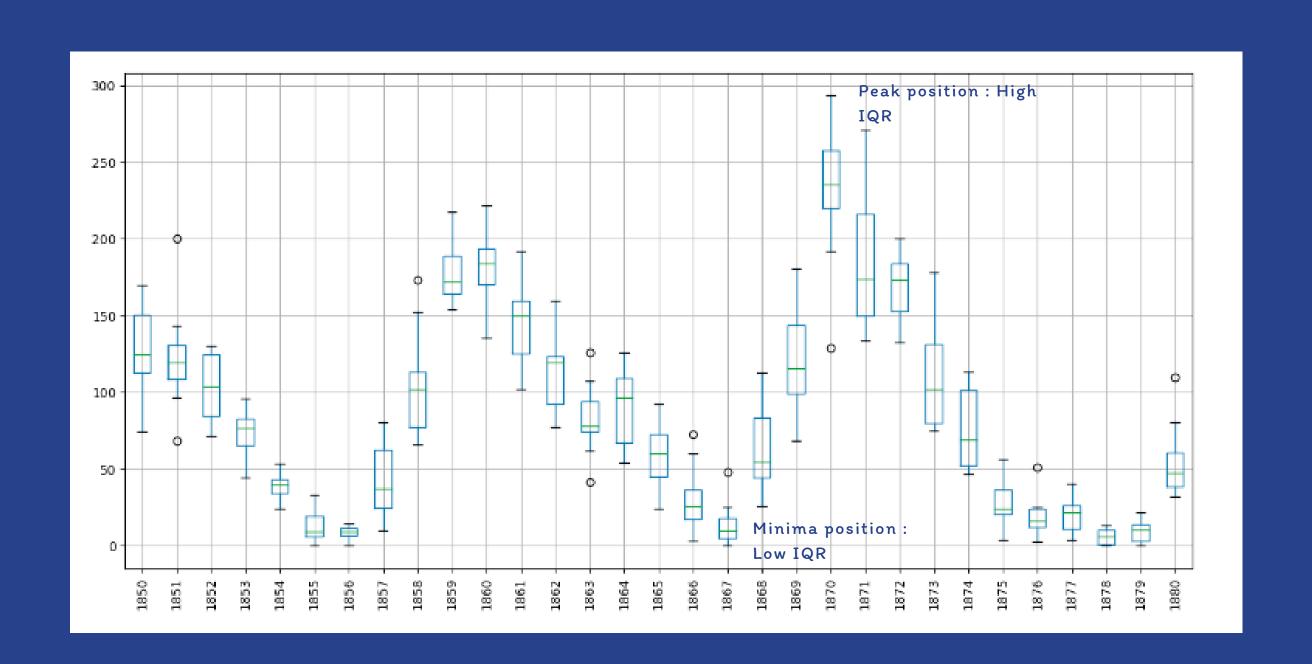








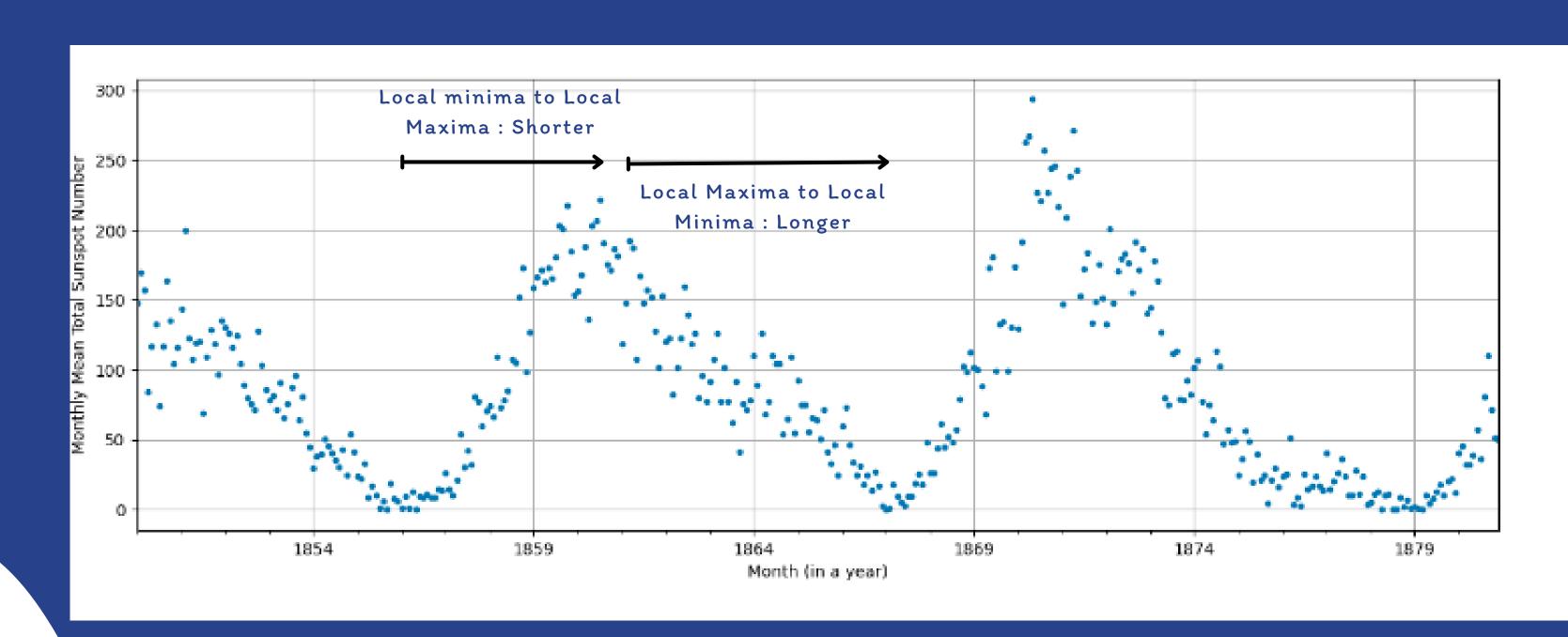
IQR OBSERVATION: CHANGES IN NUMBER OF SUNSPOTS





ASYMMETRY OF THE CHANGES

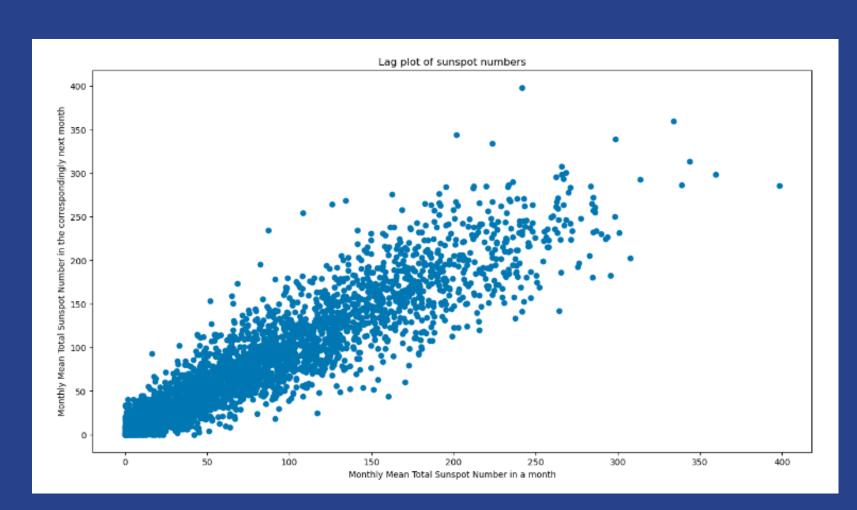


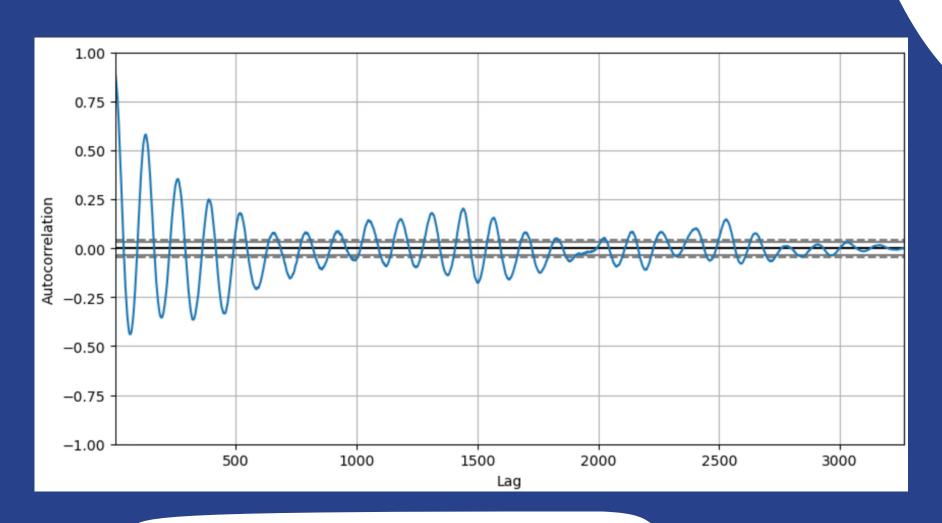






NEW CONCEPT: LAG PLOT





WHAT IS LAG PLOT?

Scatter plot with the 2 variables lagged (fixed amount of passing time)

WHAT DOES IT IMPLY?

if there is an identifiable structure in the plot generated, the data is not random.



CORE ANALYSIS: PERIOD OF THE CHANGE

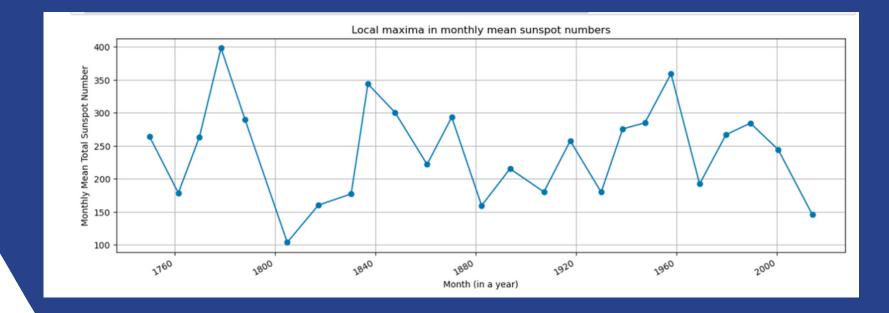
IDEA?

Find the time difference between the two adjacent local maximas, find the average.

FIRST APPROACH: FINDING LAG VALUE

```
In [77]: lag_vals = np.arange(12*15)
    autocorr_ser = np.array([sunspot_data['Monthly Mean Total Sunspot Number'].autocorr(lag=n) for n in lag_vals])
    argrelextrema(autocorr_ser, np.greater, order = 2)[0]
Out[77]: array([128], dtype=int64)
```

SECOND APPROACH: PLOTTING THE PEAK ACTIVITY FOR EACH PERIOD, FIND THE DIFFERENCE IN TIME

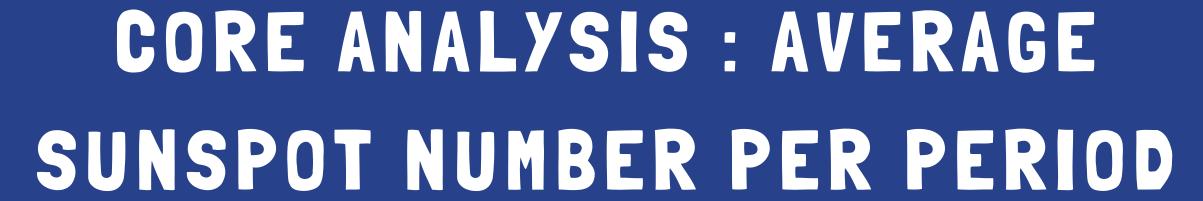


7 13.0
8 6.0
9 11.0
10 13.0
11 10.0
12 12.0
13 11.0
14 14.0
15 10.0
16 12.0
17 9.0
18 9.0
19 10.0
20 12.0
21 10.0
22 10.0
23 11.0
24 14.0
Name: Date, dtype: float64

count 24.000000 11.041667 mean std 2.330873 min 6.000000 25% 9.750000 50% 11.000000 12.250000 17.000000 max Name: Date, dtype: float64

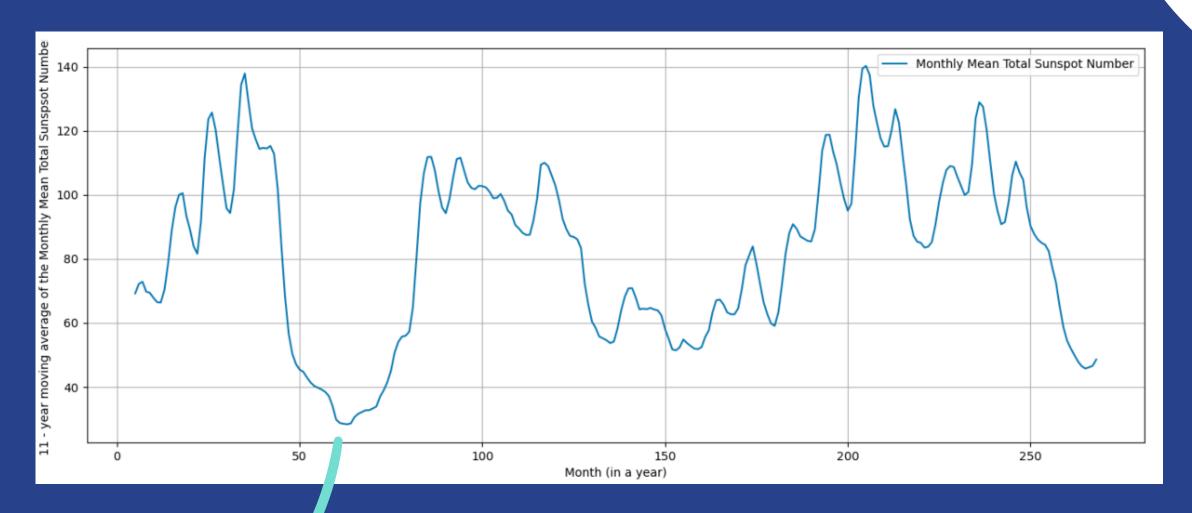
PERIOD: 11.04 YEARS





NEW CONCEPT: ROLLING AVERAGE

Rolling average: Find the average for the previous
11 years to find the average sunspot numbers.

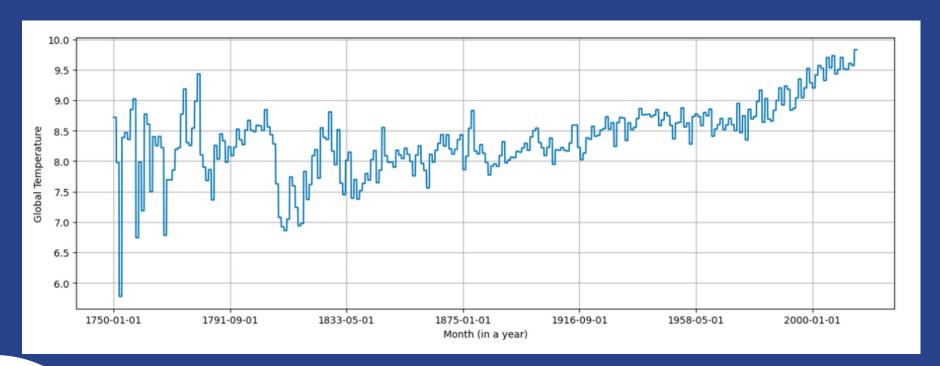


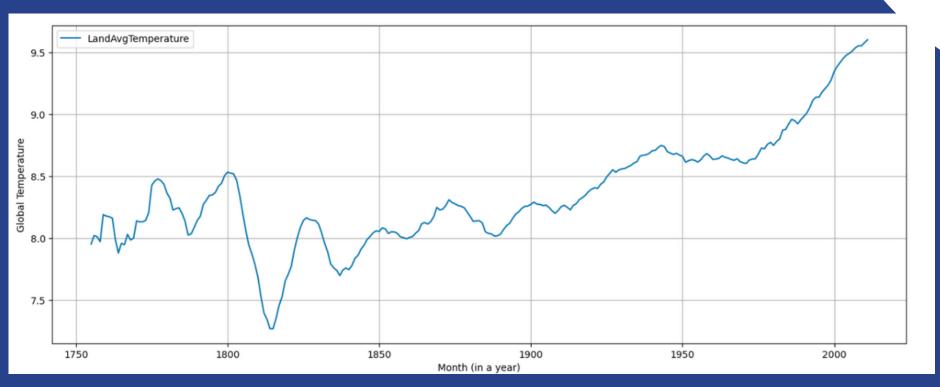
During the year 1790-1820, the solar activity reached its peak minimum: Dalton Minimum

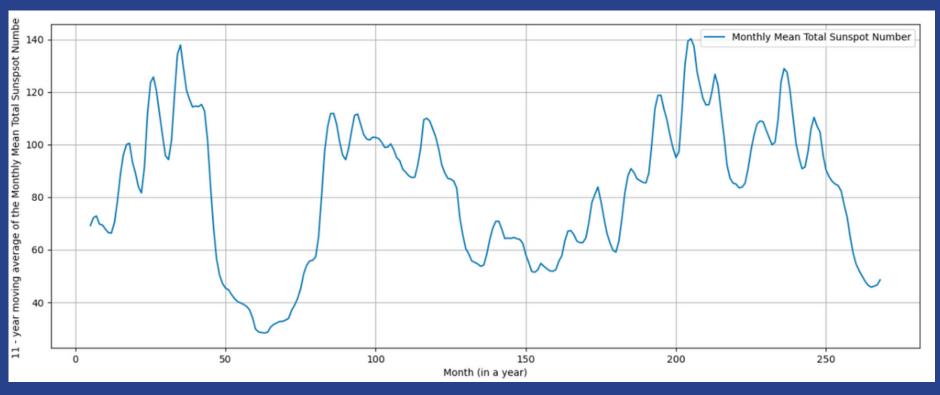


Comparison

Global Temperature Data:



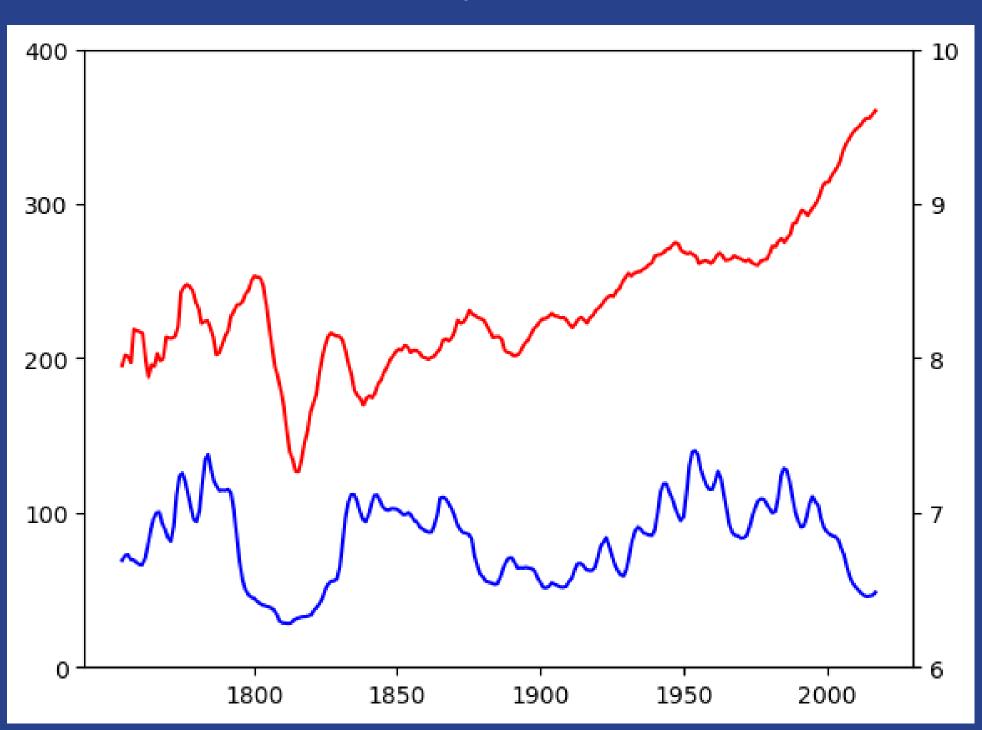






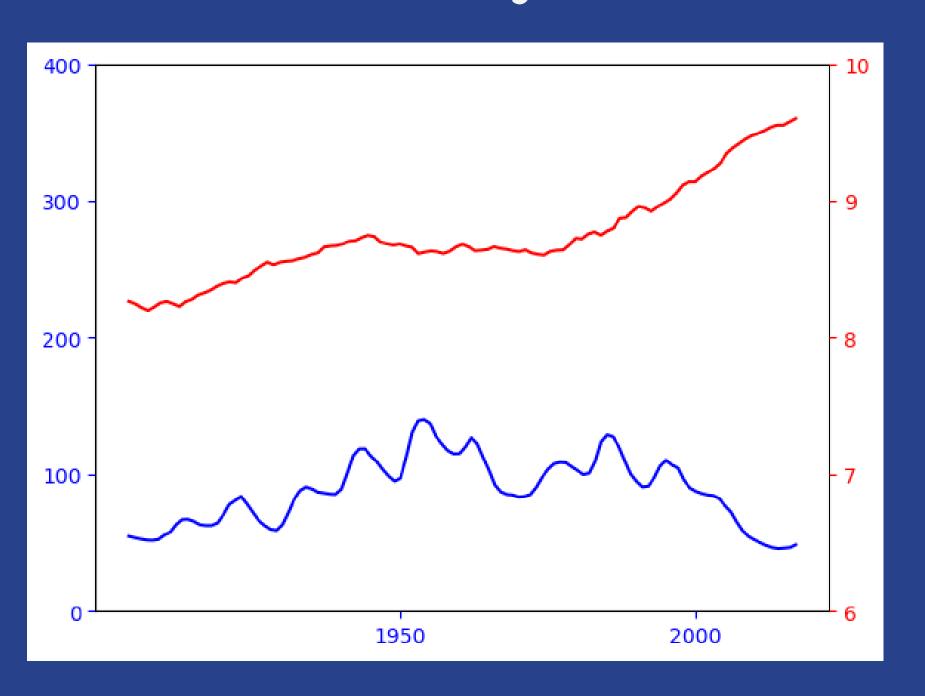


Comparison

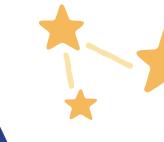




Dissimilar trend starting from 1960 - 2011

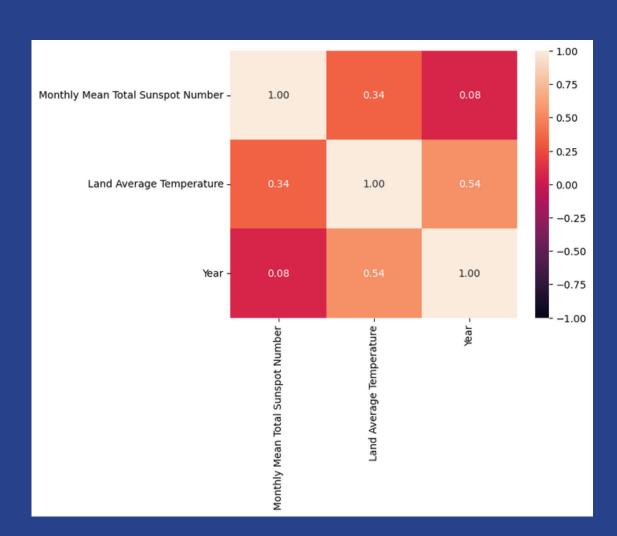






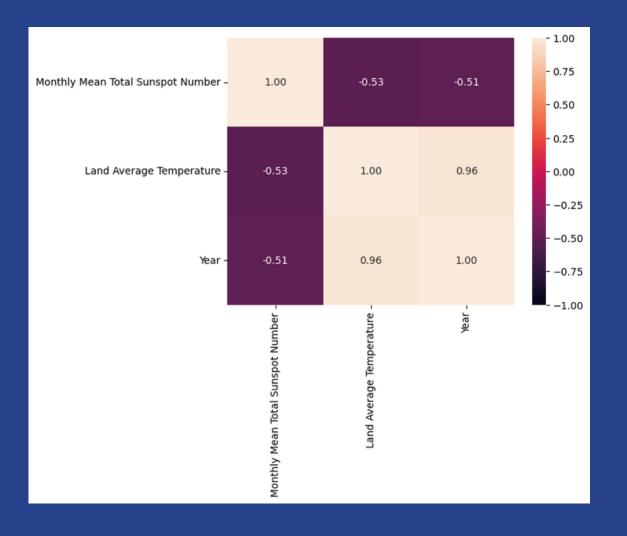
Correlation between Sunspot and Global Temperature

1755-1960



Correlation Cefficient: 0.34

1960 - 2011

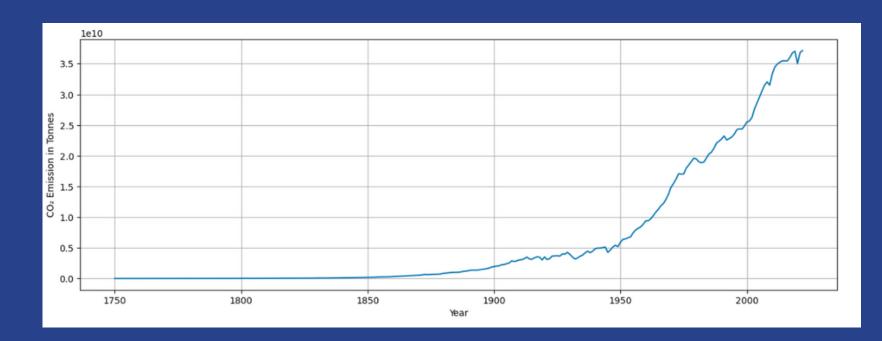


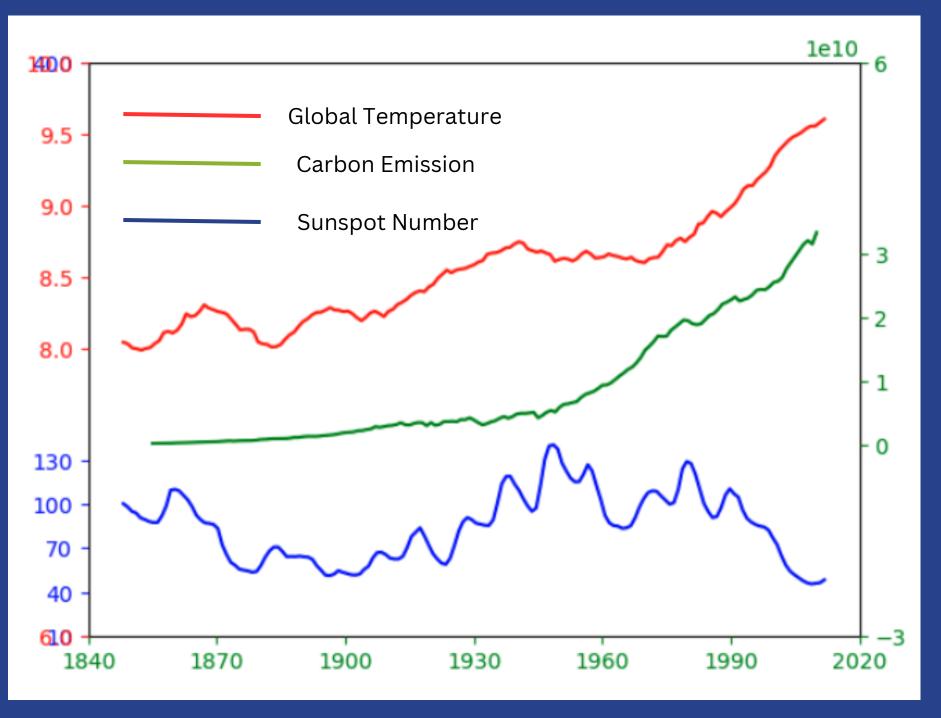
Correlation Cefficient : -0.51



CORE ANALYSIS: CORRELATION BETWEEN GLOBAL TEMPERTURE AND GLOBAL CARBON EMISSION

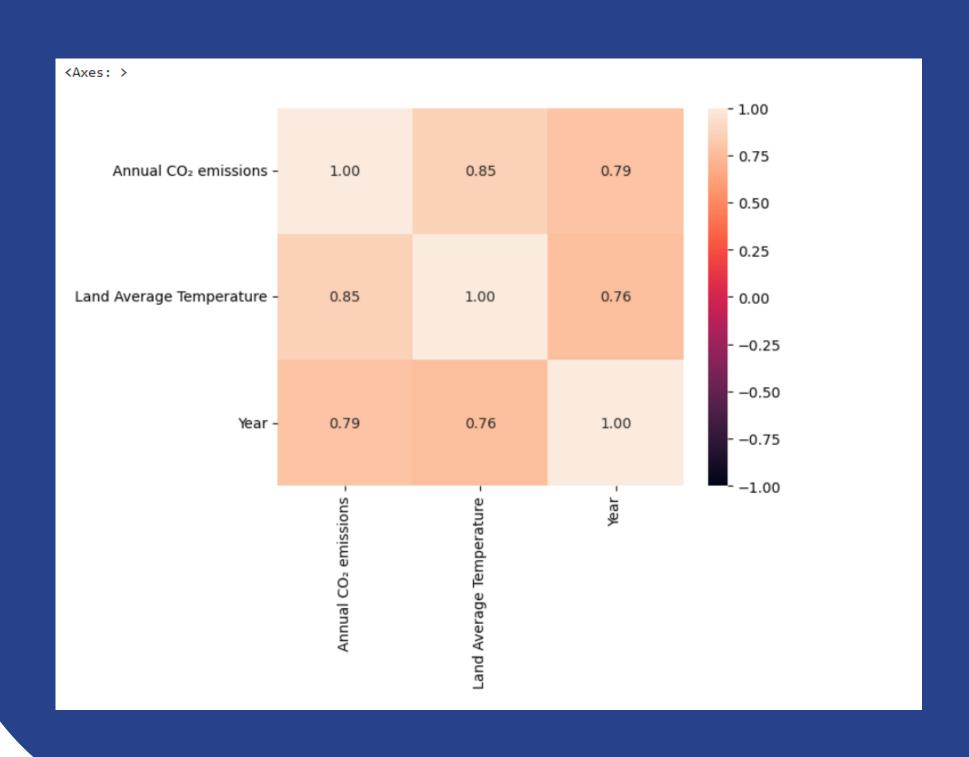
Global Carbon Emission Data:







CORE ANALYSIS: CORRELATION BETWEEN GLOBAL TEMPERTURE AND GLOBAL CARBON EMISSION

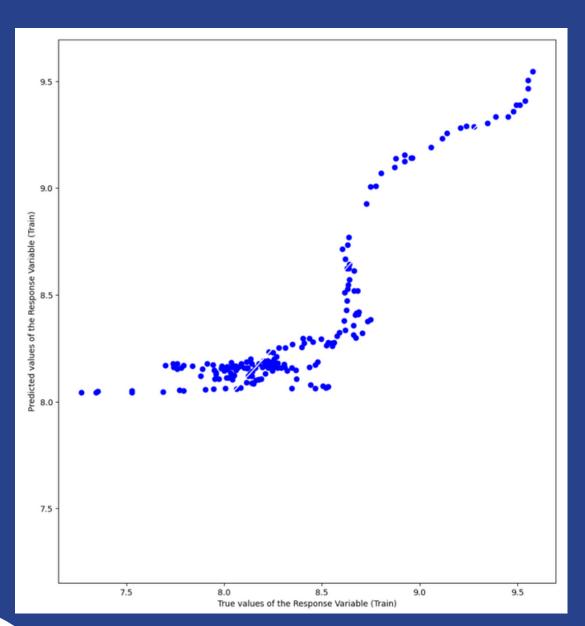


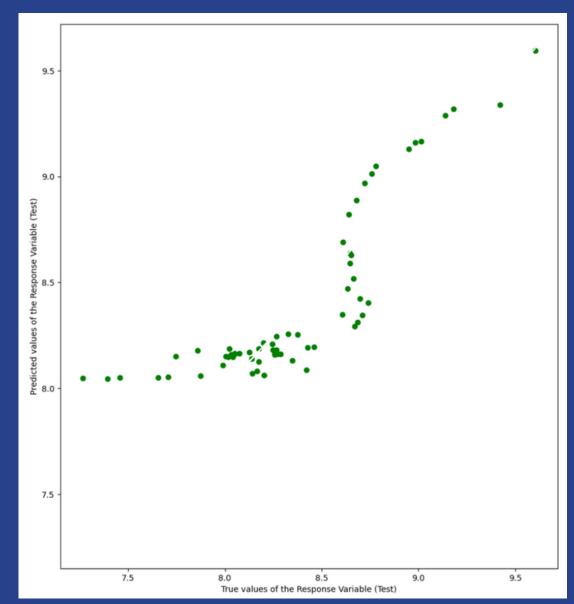
Correlation: 0.85
Implies strong correlation



MULTIVARIATE REGRESSION







Goodness of Fit of Model Train Dataset
Explained Variance (R^2) : 0.7555984655445439
Mean Squared Error (MSE) : 0.04776610847209533

Goodness of Fit of Model Test Dataset
Explained Variance (R^2) : 0.720716427

Explained Variance (R^2) : 0.7207164277356004 Mean Squared Error (MSE) : 0.05637625102464322

Global temperature = 1.5692e-03 * (monthly mean total sunspot number) + 4.9446e-11 * (carbon emission) + 7.9979

