Mini Project - Data Science and Big Data Analytics - Google Trend Analysis

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Importing Packages Required

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
    # Matplotlib Inline command is a magic command that makes the plots
    # generated by matplotlib show into the IPython shell that we are
    # running and not in a separate output window.
sns.set()
```

```
In [3]: data = pd.read_csv('multiTimeline.csv', skiprows = 1)
# add the skiprows argument to skip the first row
# at the start of the file
data.head()
```

Out[3]:

	Week	Virat Kohli: (India)	Rohit Sharma: (India)	MS Dhoni: (India)	Lionel Messi: (India)	Cristiano Ronaldo: (India)
0	2018-05- 13	14	4	11	4	5
1	2018-05- 20	15	4	16	5	7
2	2018-05- 27	11	3	21	7	9
3	2018-06- 03	11	3	8	8	7
4	2018-06- 10	10	2	8	17	21

In [6]: # add .info() method to look upon the data types of all the rows
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 260 entries, 0 to 259
Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype
			
0	Week	260 non-null	object
1	Virat Kohli: (India)	260 non-null	int64
2	Rohit Sharma: (India)	260 non-null	int64
3	MS Dhoni: (India)	260 non-null	int64
4	Lionel Messi: (India)	260 non-null	int64
5	Cristiano Ronaldo: (India)	260 non-null	int64

dtypes: int64(5), object(1)
memory usage: 12.3+ KB

In [9]: # now making the data columns more presentable
data.columns = ['Week', 'Kohli', 'Rohit', 'Dhoni', 'Messi', 'Ronaldo']
data.head()

Out[9]:

	Week	Kohli	Rohit	Dhoni	Messi	Ronaldo
0	2018-05-13	14	4	11	4	5
1	2018-05-20	15	4	16	5	7
2	2018-05-27	11	3	21	7	9
3	2018-06-03	11	3	8	8	7
4	2018-06-10	10	2	8	17	21

In [12]: data.Week = pd.to_datetime(data.Week)
 data.set_index("Week", inplace = True)

In [13]: data.head()

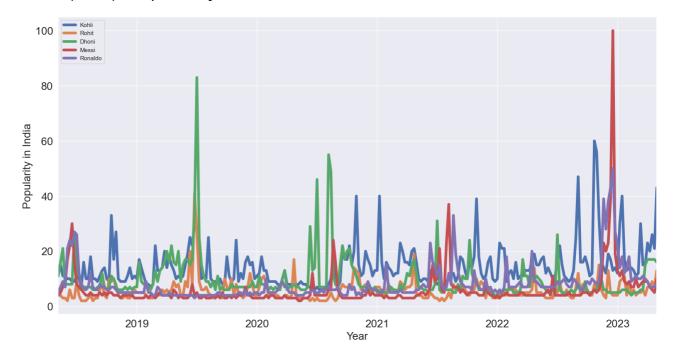
Out[13]:

Kohli Rohit Dhoni Messi Ronaldo

Week					
2018-05-13	14	4	11	4	5
2018-05-20	15	4	16	5	7
2018-05-27	11	3	21	7	9
2018-06-03	11	3	8	8	7
2018-06-10	10	2	8	17	21

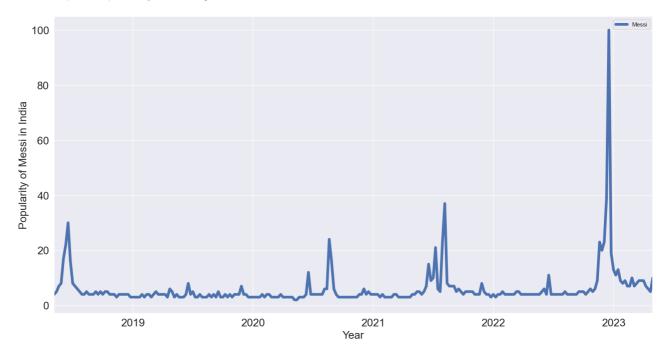
```
In [19]: data.plot(figsize=(20,10), linewidth=5, fontsize=20)
   plt.xlabel('Year', fontsize=20);
   plt.ylabel('Popularity in India', fontsize=20)
```

Out[19]: Text(0, 0.5, 'Popularity in India')



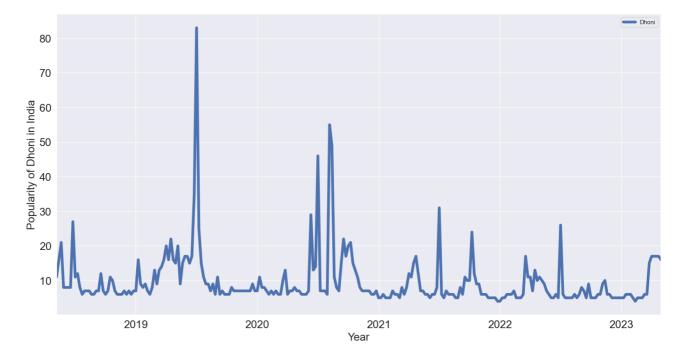
In [20]: data[['Messi']].plot(figsize=(20, 10), linewidth=5, fontsize=20)
plt.xlabel('Year', fontsize=20);
plt.ylabel('Popularity of Messi in India', fontsize=20)

Out[20]: Text(0, 0.5, 'Popularity of Messi in India')



```
In [22]: data[['Dhoni']].plot(figsize=(20, 10), linewidth=5, fontsize=20)
plt.xlabel('Year', fontsize=20);
plt.ylabel('Popularity of Dhoni in India', fontsize=20)
```

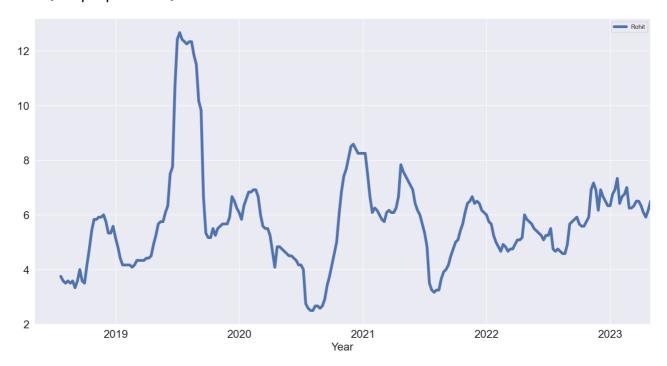
Out[22]: Text(0, 0.5, 'Popularity of Dhoni in India')



As we can see for MS Dhoni in the middle of 2019 and 2020 there is a massive spike, because of the decision of his retirement :(

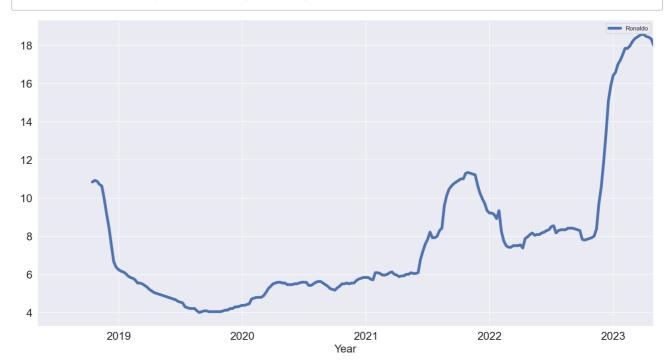
```
In [24]: rohit = data[['Rohit']]
rohit.rolling(12).mean().plot(figsize=(20, 10), linewidth=5, fontsize=20)
plt.xlabel("Year", fontsize=20)
```

Out[24]: Text(0.5, 0, 'Year')

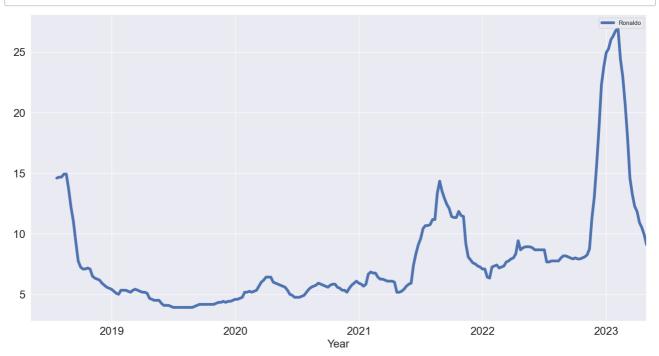


In [26]: # the method above used is done to smooth the trend analysis # by taking the average of the data and creating a specific window # here it is of 12 months # This method is also known as "Rolling Average"

In [30]: ronaldo = data[['Ronaldo']]
 ronaldo.rolling(24).mean().plot(figsize=(20, 10), linewidth=5, fontsize=20)
 plt.xlabel("Year", fontsize=20);
24 months average - rolling average method



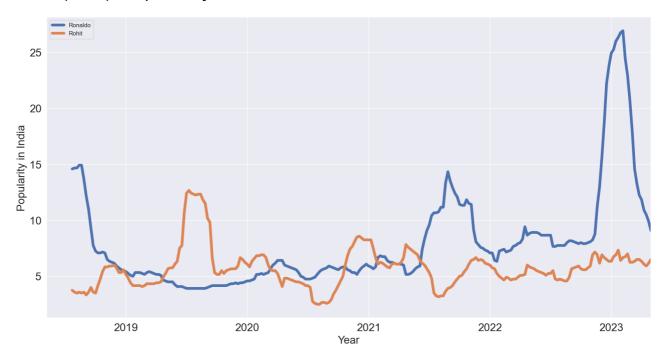
In [31]: ronaldo = data[['Ronaldo']]
 ronaldo.rolling(12).mean().plot(figsize=(20, 10), linewidth=5, fontsize=20)
 plt.xlabel("Year", fontsize=20);
12 months average - rolling average method



Now if we have to compare 2 or more attributes :

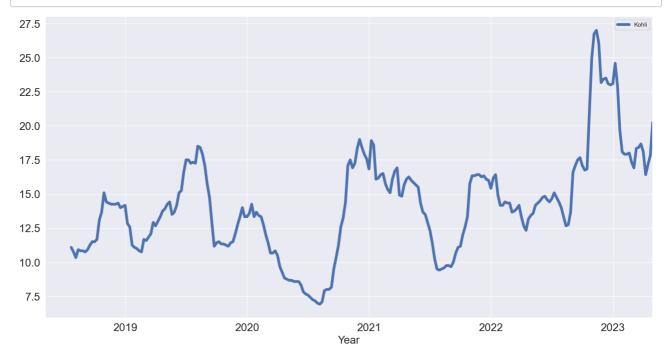
In [34]: comp = pd.concat([ronaldo.rolling(12).mean(), rohit.rolling(12).mean()], axis=1
 comp.plot(figsize=(20, 10), linewidth=5, fontsize=20)
 plt.xlabel("Year", fontsize=20)
 plt.ylabel('Popularity in India', fontsize=20)

Out[34]: Text(0, 0.5, 'Popularity in India')



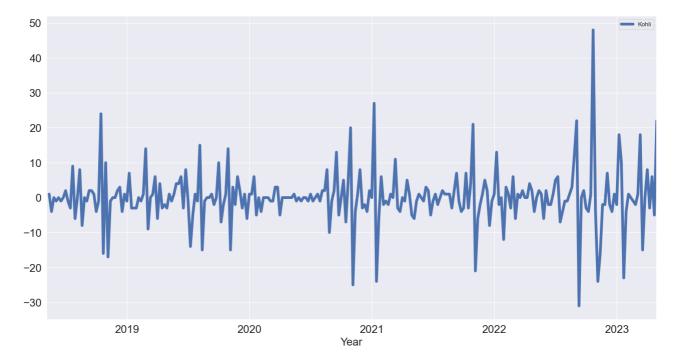
Seasonal Trends and Patterns

In [35]: virat = data[['Kohli']]
virat.rolling(12).mean().plot(figsize=(20,10), linewidth=5, fontsize=20)
plt.xlabel('Year', fontsize=20);



In [37]: virat.diff().plot(figsize=(20,10), linewidth=5, fontsize=20)
plt.xlabel("Year", fontsize=20)

Out[37]: Text(0.5, 0, 'Year')



As you can see in the seasonal trends as soon as the big cricket event comes Virat Kohli's Popularity increases.

The massive spike before 2023 was due to 71st century he scored in Asia Cup.

Correlation

In [38]: data.corr()

Out[38]:

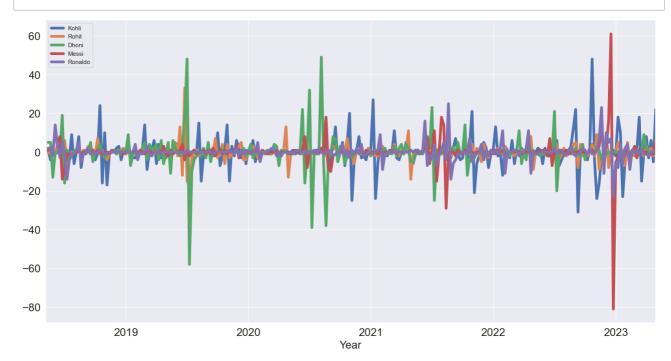
	Kohli	Rohit	Dhoni	Messi	Ronaldo
Kohli	1.000000	0.491038	0.108377	-0.004534	0.018491
Rohit	0.491038	1.000000	0.342616	-0.075430	-0.050410
Dhoni	0.108377	0.342616	1.000000	-0.058894	-0.126039
Messi	-0.004534	-0.075430	-0.058894	1.000000	0.727348
Ronaldo	0.018491	-0.050410	-0.126039	0.727348	1.000000

Virat Kohli and Rohit Sharma have a correlation of 0.49 as they play the country.

But the correlation between Virat and Messi is negative as they play in 2 different sports

Differences

In [39]: data.diff().plot(figsize=(20, 10), linewidth=5, fontsize=20)
plt.xlabel("Year", fontsize=20);



Now the combination of correlation and differences

In [40]: data.diff().corr()

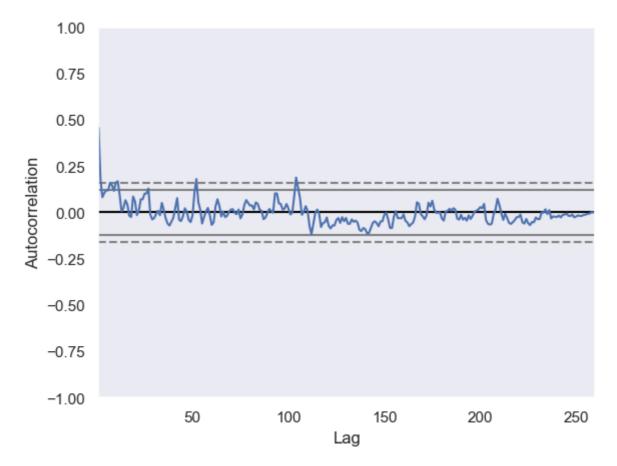
Out[40]:

	Kohli	Rohit	Dhoni	Messi	Ronaldo
Kohli	1.000000	0.289733	0.099048	0.004458	-0.022802
Rohit	0.289733	1.000000	0.118266	-0.053738	0.006081
Dhoni	0.099048	0.118266	1.000000	-0.065611	-0.002000
Messi	0.004458	-0.053738	-0.065611	1.000000	0.432781
Ronaldo	-0.022802	0.006081	-0.002000	0.432781	1.000000

Auto correlation

In [41]: pd.plotting.autocorrelation_plot(virat)

Out[41]: <AxesSubplot:xlabel='Lag', ylabel='Autocorrelation'>



Conclusion: In this project, learning that are covered are correlation, plotting, multiple value plotting, auto-correlation, differences etc.

In []:	