

# CSE485 - Data Mining

## Final Project

### “Terrorism and Ideology Analysis”

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## Abstract:

There are two datasets, "Global Terrorism Database" and "Government Composition". The links between political tendencies of ruling parties, governments and terrorism will be researched. There are 5 main countries with terrorism and government ideology data.

- US, UK, Germany, France, Spain

## Introduction and Problem:

Global Terrorism Database is one of the most analyzed databases around Kaggle platform. It includes all terrorist attacks between 1970-2017, terrorist attack's country, date, attack type, number of wounded persons, terrorist organisation identifiers etc. It's useful for exploring places of attacks.

Government Composition database holds ideology, distribution of parliament chairs, dates etc. That dataset is helpful for describing a government with its ruling years and ideology.

Main hypothesis is **If the ruler parties are tend to be "right" ideology, terrorist events are more than the other times.**

## Datasets:

- Global Terrorism Database, CSV, 155MB , <https://www.kaggle.com/START-UMD/gtd>

Government Composition Dataset, XLSX, 1.54MB, Armingeon, Klaus, Virginia Wenger, Fiona Wiedemeier, Christian Isler, Laura Knöpfel and David Weisstanner. 2018. Supplement to the Comparative Political Data Set – Government Composition 1960-2016, Bern: Institute of Political Science, University of Berne.

## Code Sources:

<https://www.kaggle.com/raghuchaudhary/global-terrorism>

<https://www.kaggle.com/ash316/terrorism-around-the-world>

## Explatory Data Analysis:

- Firstly without doing anything on datasets, I imported them and described their mean, std, count etc.

	Year	Month	Day	Country	Region	city	latitude	longitude	AttackType	Killed	Wounded	Target	Summary	Group
0	1970	7	2	Dominican Republic	Central America & Caribbean	Santo Domingo	18.456792	-69.951164	Assassination	1.0	0.0	Julio Guzman	NaN	MANO-D
1	1970	0	0	Mexico	North America	Mexico city	19.371887	-99.086624	Hostage Taking (Kidnapping)	0.0	0.0	Nadine Chaval, daughter	NaN	23rd of September Communist League
2	1970	1	0	Philippines	Southeast Asia	Unknown	15.478598	120.599741	Assassination	1.0	0.0	Employee	NaN	Unknown
3	1970	1	0	Greece	Western Europe	Athens	37.997490	23.762728	Bombing/Explosion	NaN	NaN	U.S. Embassy	NaN	Unknown
4	1970	1	0	Japan	East Asia	Fukouka	33.580412	130.396361	Facility/Infrastructure Attack	NaN	NaN	U.S. Consulate	NaN	Unknown
5	1970	1	1	United States	North America	Cairo	37.005105	-89.176269	Armed Assault	0.0	0.0	Cairo Police Headquarters	1/1/1970: Unknown African American assailants ...	Black Nationalists
6	1970	1	2	Uruguay	South America	Montevideo	-34.891151	-56.187214	Assassination	0.0	0.0	Juan Maria de Lucah/Chief of Directorate of in...	NaN	Tupamaros (Uruguay)
7	1970	1	2	United	North	Oakland	37.791927	122.225006	Bombing/Explosion	0.0	0.0	Edes	1/2/1970: Unknown namestrans	Unknown

**Fig. 1.0**

The type and first 8 rows of global terrorism data can be clearly seen in **Fig. 1.0**.

I will use only Year and Country. Because every row shows a different attack with different features.

## Describing of GTDB

	Year	Month	Day	latitude	longitude	Killed	Wounded	casualties
count	181691.000000	181691.000000	181691.000000	177135.000000	1.771340e+05	171378.000000	165380.000000	164817.000000
mean	2002.638997	6.467277	15.505644	23.498343	-4.586957e+02	2.403272	3.167668	5.296128
std	13.259430	3.388303	8.814045	18.569242	2.047790e+05	11.545741	35.949392	42.069023
min	1970.000000	0.000000	0.000000	-53.154613	-8.618590e+07	0.000000	0.000000	0.000000
25%	1991.000000	4.000000	8.000000	11.510046	4.545640e+00	0.000000	0.000000	0.000000
50%	2009.000000	6.000000	15.000000	31.467463	4.324651e+01	0.000000	0.000000	1.000000
75%	2014.000000	9.000000	23.000000	34.685087	6.871033e+01	2.000000	2.000000	4.000000
max	2017.000000	12.000000	31.000000	74.633553	1.793667e+02	1570.000000	8191.000000	9574.000000

**Fig 1.1**

The first impression on untouched data is quite interesting.

What can we say from looking to Fig 1.1?

- There were 181.691 Terrorist attacks happened between 1970-2017
  - Attacks were happened around June, 15<sup>th</sup> day.
  - 2.4 People died, 3.16 people wounded per attack.
  - Maximum kills in a single attack is 1570, max woundeds 8191.
- 
- And then, let's describe governments data.

	Year	Country	Ideology
0	1959.000000	USA	right
189	1992.000000	UK	right
175	1985.000000	UK	right
176	1986.000000	UK	right
177	1986.000000	UK	right
178	1987.000000	UK	right
179	1987.000000	UK	right
180	1988.000000	UK	right
181	1988.000000	UK	right
182	1989.000000	UK	right
183	1989.000000	UK	right
184	1990.000000	UK	right

Fig 1.3

As it seen, only year, country and ideology of government data used in Government Composition Database (GCDB).

- Exploring terrorist attacks depends on years.

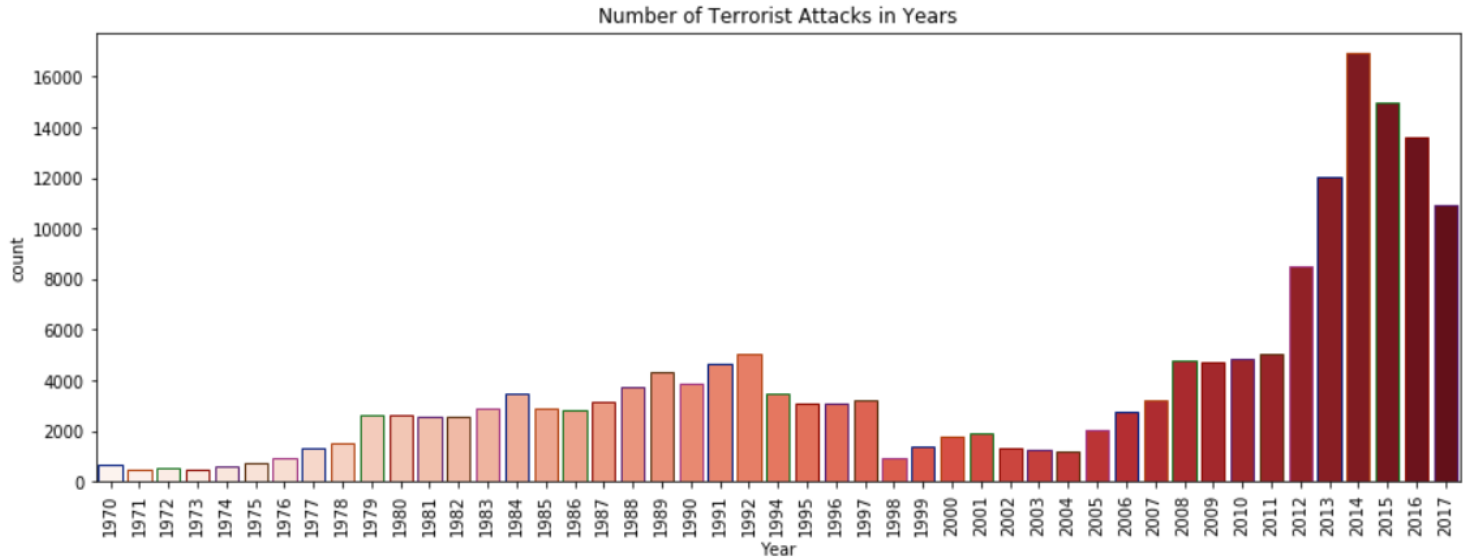


Fig 2.0

The year which has most terrorist attacks happened is 2014. Figure 2.0 shows attacks increased after 2004.

- Governments analysis.

We should look at the number of ideologies governments has support.

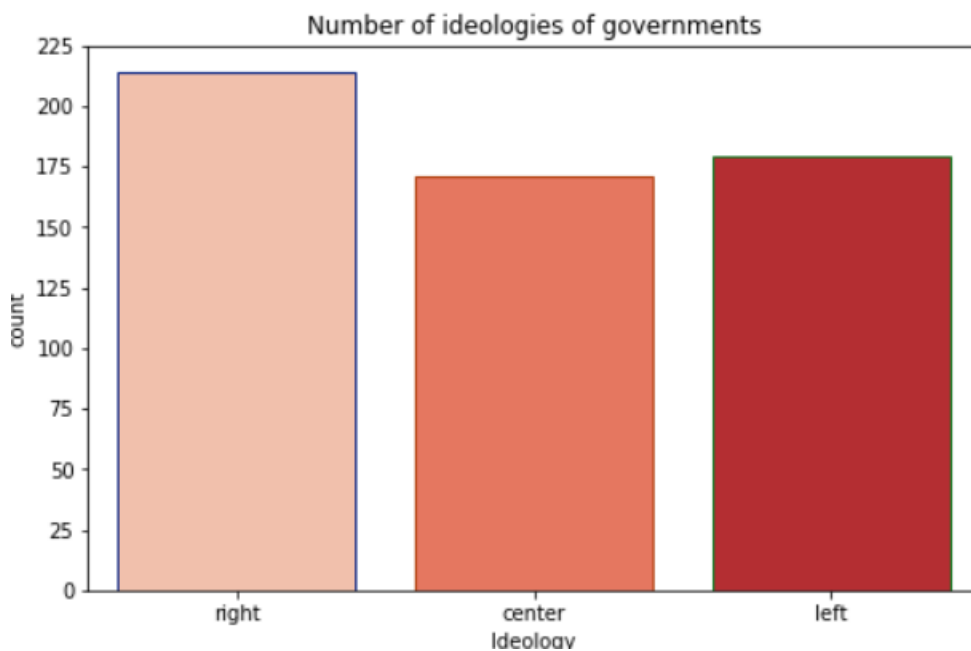


Fig 2.1

Fig. 2.1 shows us the number of ideologies that governments have. Between 1960-2016, Right Ideology is the most supported.

- Now, get these two results together. If these two datas gets together depends on their “Year” and “Country” column with pandas.concat(), all the terrorist attack in our five country will be listed.

	Country	Ideology	Year
0	Dominican Republic	NaN	1970.0
1	Mexico	NaN	1970.0
2	Philippines	NaN	1970.0
3	Greece	NaN	1970.0
4	Japan	NaN	1970.0
5	United States	NaN	1970.0
6	Uruguay	NaN	1970.0
7	United States	NaN	1970.0
8	United States	NaN	1970.0
9	United States	NaN	1970.0
10	United States	NaN	1970.0
11	United States	NaN	1970.0
12	Italy	NaN	1970.0
13	United States	NaN	1970.0
14	United States	NaN	1970.0
15	East Germany (GDR)	NaN	1970.0
16	Ethiopia	NaN	1970.0
17	United States	NaN	1970.0
18	United States	NaN	1970.0

After concatenating our two lists, there are so many Countries which are not useful for this data, and null fields.

All of them have to be cleaned.

Fig 3.0

19	1972	France	right
20	1972	France	right
21	1972	France	right
22	1973	France	right
23	1973	France	right
24	1973	France	right
25	1973	France	right
26	1973	France	right
27	1973	France	right
28	1973	France	right
29	1973	France	right
...	...	...	...
12936	2016	Germany	center
12937	2016	Germany	center
12938	2016	Germany	center
12939	2016	Germany	center
12940	2016	Germany	center
12941	2016	Germany	center
12942	2016	Germany	center

Remaining dataset is so clear and useful to understand.

Fig 3.1

- Finally, if we visualize that data depends on It's Ideology column, the main hypothesis can be tested.

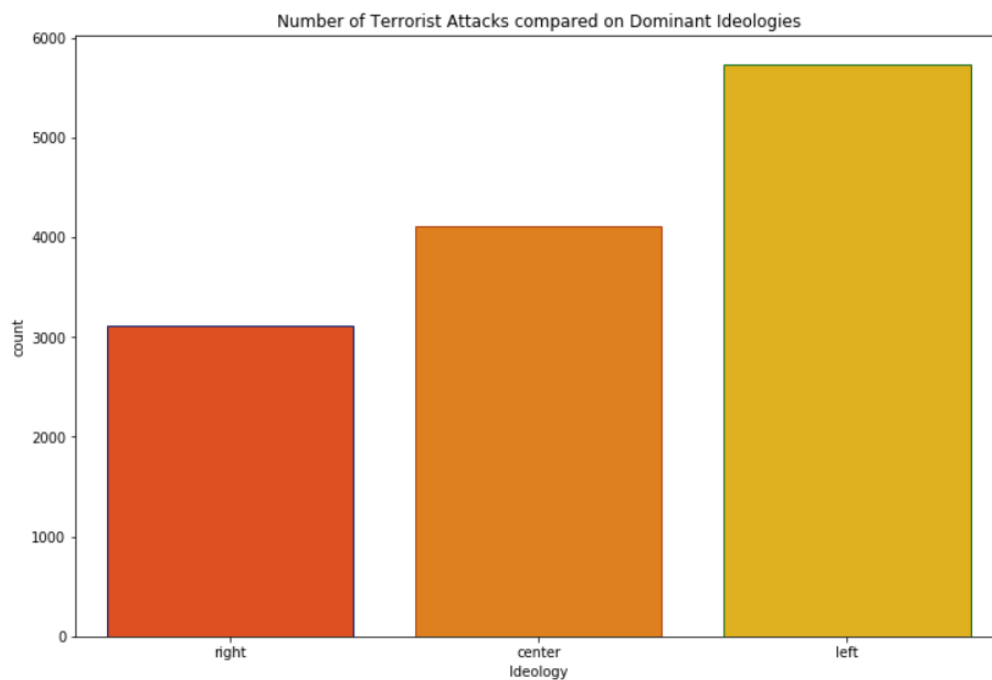


Fig 4.0

- Fig 4.0 shows that hypothesis is totally opposite for 5 countries. Terrorist attacks happening while left ideology is on rule more than right ideology.
- 5 Clusters K-Means test on final data.

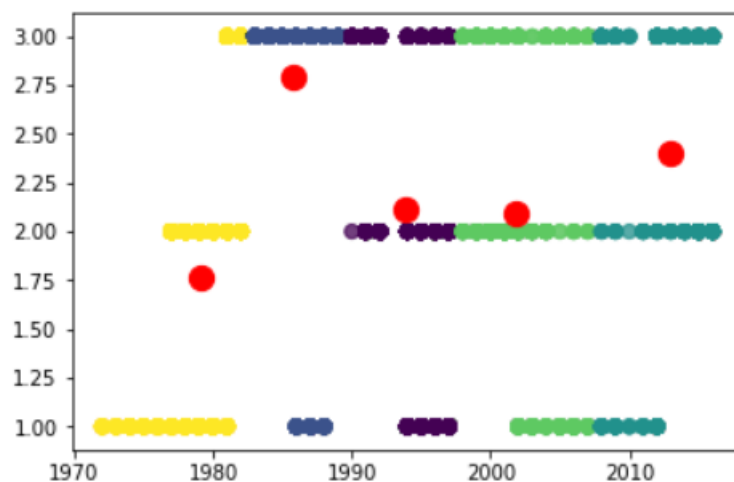


Fig 4.1

## Conclusion:

There are several final results in this analysis.

- 1- Governments which has left ideology are weak on counterterrorism.
- 2- Governments which has left ideology may be founding terrorist organisations.
- 3- Governments which has Right ideology are better than left ones on counterterrorism.
- 4- Governments which has on center ideology are not successful against terrorist organisations.



After all these results, It is cleary open that main hypothesis for that analysis is false.

The opposite of the main hypothesis has been proven.