Exploring the Connection between Terrorism and Freedom Indices

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Abstract

Ever since 9/11 terrorism has become synonymous with transnational terrorsim in popular media. A terrorist has often times become equal to a Muslim. Furthermore, the popular public perception of terrorism is the one of opression. This research project aims to provide broad overview of that. It does not explore this complicated relationship in detail, but rather provides broad overview of how the available data on terrorist attacks correlates with different indices for freedom. We specifically look into freedom of expression, religion, movement and conflicts in general.

Platform where the application runs: NYU Dumbo cluster

Introduction

Motivation

- We have lived in a Muslim country for the past 3 years and get asked a lot about terrorism in relation to Islam and the lack of freedom. Hence, we wanted to **explore for ourselves how terrorism relates to the various indices of freedom**.
- There has been a lot of research done with regards to transnational terrorism, but we were not able to find substantial data on how terrorism relates to freedom in general.

Users

• **General public** to inform themselves of the relations, **popular media** to inform themselves when representing terrorism, **researchers** to investigate this more in depth than we were able to.

Introduction

Beneficiaries

• **General Public** to explore implicit biases we hold when it comes to talking about terrorism.

Importance

- We show that there is weak correlation between religion and the acts of terrorism, which is important given the public perceptions of terrorism.
- We show that terrorism is mostly correlated with the conflict itself, which is supported by the existing research.
- Our results support the claim that terrorism is hard to define, hence, we cannot understand it simply as the loss of lives.
- We show that there is a very weak correlation between freedom of press (sign of democratic and non-opressive governments) and acts of terrorism, which go against the popular perceptions of terrorism existing in oppresive regimes.

Goodness

- We took time in understanding the components of the datasets and the domain insights to frame the various hypotheses.
- Given the multi-dimensional nature of the topic, we made sure to explore multitudes of different factors to gain a fuller insight of the topic.
- We referred to journals and articles to extend the analytics and to provide a context to each results.
- Took extra care in bridging the two datasets in terms of matching up the data and provide integrity in the combined data.

Data Sources

Global Terrorism Database (GTD) 1970-2018 | 173.5 MB

- A comprehensive dataset on terrorist attacks in the world by Global National Consortium for the Study of Terrorism and Responses to Terrorism for 1970-2018.
- https://gtd.terrorismdata.com/files/gtd-1970-2018/

Human Freedom Index (HFI) 2019 | 672 B

- A dataset on human freedom in the world, using 76 distinct indicators of personal and economic freedom, of 2008-2017.
- https://www.cato.org/human-freedom-index-new

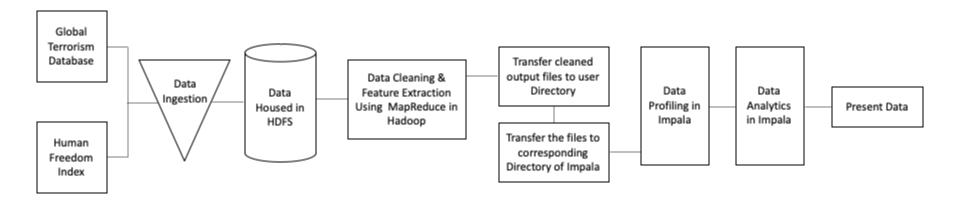
Data Sample #1: GTD 1970-2018

·		+		+	+	
event_id year mont attack type	n day country region	nationality	doubt_terror		ss suici	Lde
	++				+	
	.+		+	+		
200801010001 2008 1	1 Philippines Southeast A	sia	0	1	0	
Armed Assault	Business	Australia	Θ	1		
200801010002 2008 1	1 Kenya Sub-Saharan	Africa	0	1	0	
Armed Assault	Religious Figures/Institutions		50			
200801010003 2008 1	1 India South Asia		0	1	0	
Armed Assault	Police	India	8	1		
200801010004 2008 1	1 Sudan Sub-Saharan		1	1	0	
Armed Assault	Government (Diplomatic)		2	1		
200801010005 2008 1	1 Nigeria Sub-Saharan		0	1	0	
Armed Assault	Police	Nigeria	18	1		
200801010006 2008 1	1 Sri Lanka South Asia		0	1	0	
Armed Assault	Government (General)		2			
200801010008 2008 1		& North Africa		1	1	
Bombing/Explosion	1	Iraq	37	1 1 1	1.0	
200801010009 2008 1 Armed Assault	1 Iraq Middle East	Iraq	1 1	1	0	
200801010010 2008 1		& North Africa	0 1	1	ΙΘ	
Assassination	Police	Iraq	1.5	1 1 1	1 0	
200801010011 2008 1		& North Africa I		1	1 0	
Armed Assault		Iraq	1 1	1	1 0	
		& North Africa	1 1	1 1 1	1.0	
Bombing/Explosion	Military	Iraq	1 1	1	1 0	
200801010013 2008 1	1 Kosovo Eastern Eur		0	1 1	1 0	
Bombing/Explosion	Business	Serbia-Montenegr	0 1 0	1	, -	
200801010014 2008 1	1 Pakistan South Asia		1	1	0	
Hostage Taking (Kidnapping)	Military	Pakistan	1 0	1	1.000	

Data Sample #2: HFI 2019

+		+	+	++
country	region	hf_score	pf_score	ef_score
+			+	++
Albania	Eastern Europe	7.679999828338623	8.069999694824219	7.28000020980835
Algeria	Middle East & North Africa	5.289999961853027	5.349999904632568	5.239999771118164
Angola	Sub-Saharan Africa	5.019999980926514	5.440000057220459	4.590000152587891
Argentina	Latin America & the Caribbean	7.099999904632568	8.189999580383301	6.010000228881836
Armenia	Caucasus & Central Asia	7.420000076293945	7.210000038146973	7.630000114440918
Australia	Oceania	8.729999542236328	9.289999961853027	8.180000305175781
Austria	Western Europe	8.420000076293945	9.020000457763672	7.820000171661377
Azerbaijan	Caucasus & Central Asia	6.440000057220459	6.730000019073486	6.150000095367432
Bahamas	Latin America & the Caribbean	7.96999979019165	8.319999694824219	7.610000133514404
Bahrain	Middle East & North Africa	7.150000095367432	6.880000114440918	7.409999847412109
Bangladesh	South Asia	5.800000190734863	5.599999904632568	6.010000228881836
Barbados	Latin America & the Caribbean	7.210000038146973	8.220000267028809	6.199999809265137
Belarus	Eastern Europe	-1	-1	-1
Belgium	Western Europe	8.279999732971191	9.060000419616699	7.5
Belize	Latin America & the Caribbean	7.369999885559082	7.400000095367432	7.329999923706055
Benin	Sub-Saharan Africa	6.849999904632568	7.460000038146973	6.239999771118164
Bhutan	South Asia	-1	-1	-1
Bolivia	Latin America & the Caribbean	6.75	7.460000038146973	6.039999961853027
Bosnia and Herzegovina	Eastern Europe	7.269999980926514	8	6.539999961853027
Botswana	Sub-Saharan Africa	6.809999942779541	6.75	6.869999885559082
+			+	++

Design Diagram



Code Challenge

When we were merging the data sets, we realized that the names of the countries did not match. To overcome this, we simply hard coded the changed names to match. We chose to use the names from the GTD dataset.

 First we found the country names that did not match, looked through them and manually paired up the names we believed belonged to the same country.

```
drop table if exists distinct_country;
create external table distinct_country (country string);
truncate table distinct_country;
insert into distinct_country select gtd_country from(select distinct gtd_country from gtd_distinct_country
    union ALL select distinct hfi_country from hfi_distinct_country) as tbl group by gtd_country having count(*) =1;
```

 Then we manually updated the HFI table to match the expressions from the GTD table.

```
table if exists hfi master1:
create external table hfi_master1 (year int,country string, region string, hf_score float, pf_rol float,pf_ss_disappearances_disap float,
   pf ss disappearances violent float.pf ss disappearances organized float.pf ss disappearances fatalities float.pf ss disappearances injuries float.
   pf ss disappearances float,pf ss women fgm float,pf ss women inheritance widows float,pf ss women inheritance daughters float,pf ss women inheritance float,
   pf ss women float,pf ss float,pf movement domestic float,pf movement foreign float,pf movement float,pf religion estop establish float,
   pf religion estop operate float,pf religion estop float,pf religion harassment float,pf religion restrictions float,pf religion float,
   pf association float.pf expression killed float.pf expression jailed float.pf expression influence float.pf expression control float.
   pf expression cable float,pf expression newspapers float, pf expression internet float,pf expression float,pf identity float,pf score float,
   ef trade tariffs revenue float, ef trade tariffs float,ef trade black float,ef trade movement foreign float,ef trade movement capital float,
   ef trade movement visit float, ef trade movement float, ef trade float, ef regulation labor float, ef regulation business float, ef regulation float,
   ef score float):
 runcate table hfi master1;
           hfi master1 (year, country, region, hf score, pf rol,pf ss disappearances disap,pf ss disappearances violent,pf ss disappearances organized,
   pf ss disappearances fatalities,pf ss disappearances injuries,pf ss disappearances,pf ss women fgm,pf ss women inheritance widows,pf ss women inheritance daughters,
   pf ss women inheritance,pf ss women,pf ss,pf movement domestic,pf movement foreign,pf movement,pf religion estop establish,pf religion estop operate,
   pf religion estop.pf religion harassment.pf religion restrictions.pf religion.pf association.pf expression killed.pf expression jailed.pf expression influence.
   pf_expression_control,pf_expression_cable,pf_expression_newspapers, pf_expression_internet,pf_expression,pf_identity,pf_score, ef_trade_tariffs_revenue,
   ef_trade_tariffs,ef_trade_black,ef_trade_movement_foreign,ef_trade_movement_capital,ef_trade_movement_visit,ef_trade_movement,ef_trade, ef_regulation_labor,
   ef regulation business, ef regulation, ef score)
    ct year, (ca
    when country='Yemen, Rep.' Then 'Yemen'
        country='Slovak Rep.' Then 'Slovak Republic'
        country='Pap. New Guinea' Then 'Papua New Guinea'
        country='Kyrgyz Republic' Then 'Kyrgyzstan'
        country='Korea, South' Then 'South Korea'
        country='Bosnia and Herzegovina' Then 'Bosnia-Herzegovina'
        country='Dominican Rep.' Then 'Dominican Republic
       country='Central Afr. Rep.' Then 'Central African Republic'
       n country='Czech Rep.' Then 'Czech Republic'
        country='Congo, Dem. R.' Then 'Democratic Republic of the Congo'
        country='Congo, Rep. Of' Then 'Republic of the Congo' else country end) as country, (case
                             en 'Western Europe' else region end) as region, hf_score, pf_rol,pf_ss_disappearances_disap,pf_ss_disappearances_violent,
   pf ss disappearances organized.pf ss disappearances fatalities.pf ss disappearances injuries.pf ss disappearances.pf ss women fgm,
   pf_ss_women_inheritance_widows,pf_ss_women_inheritance_daughters,pf_ss_women_inheritance,pf_ss_women,pf_ss,pf_movement_domestic,
   pf movement foreign.pf movement.pf religion estop establish.pf religion estop operate.pf religion estop.pf religion harassment.pf religion restrictions.
   pf religion,pf association,pf expression killed,pf expression jailed,pf expression influence,pf expression control,pf expression cable,pf expression newspapers,
    pf expression internet,pf expression,pf identity,pf score, ef trade tariffs revenue, ef trade tariffs,ef trade black,ef trade movement foreign,ef trade movement capital,
     f_trade_movement_visit,ef_trade_movement,ef_trade, ef_regulation_labor, ef_regulation_business,ef_regulation,ef_score from hfi_raw;
```

 Lastly, we made an intersection table, which we then used that make 'master' tables hfi_master and gtd_master by checking if the year, country combination exists in this intersection table.

```
drop table if exists cy intersection;
 create table cy intersection (year int, country string, region string, year2 int, country2 string, region2 string);
truncate table cy intersection;
 insert into cy intersection select * from cy intersection gtd inner join cy intersection hfi on cy intersection gtd.year=cy intersection hfi.year
     and cy intersection gtd.country=cy intersection hfi.country and cy intersection gtd.region=cy intersection hfi.region;
 rop table if exists hfi master;
create external table hfi_master(year int,country string, region string, hf_score float, pf_rol float,pf_ss_disappearances_disap float,
   pf ss disappearances violent float,pf ss disappearances organized float,pf ss disappearances fatalities float,pf ss disappearances injuries float,
   pf ss disappearances float,pf ss women fgm float,pf ss women inheritance widows float,pf ss women inheritance daughters float,pf ss women inheritance float,
   pf ss women float,pf ss float,pf movement domestic float,pf movement foreign float,pf movement float,pf religion estop establish float,
   pf_religion_estop_operate float,pf_religion_estop float,pf_religion_harassment float,pf_religion_restrictions float,pf_religion float,
   pf association float,pf expression killed float,pf expression jailed float,pf expression influence float,pf expression control float,
   pf expression cable float,pf expression newspapers float, pf expression internet float,pf expression float,pf identity float,pf score float,
   ef trade tariffs revenue float, ef trade tariffs float,ef trade black float,ef trade movement foreign float,ef trade movement capital float,
   ef trade movement visit float, ef trade movement float, ef trade float, ef regulation labor float, ef regulation business float, ef regulation float, ef score float);
truncate table hfi master;
 insert into hfi master select * from hfi master1 where EXISTS (select year, country FROM cy intersection WHERE hfi master1.year=cy intersection.year
   and hfi master1.country=cy intersection.country);
```

Furthermore, the region names also did not match. GTD had more regions listed than HFI, which is why we decided to merge some of GTD ones to match the ones in HFI.

```
drop table if exists gtd_master1;
create external table gtd_master1 (event_id string, year int, month int, day int, country string, region string, doubt_terrorist int,
    success int, suicide int, attack_type string, target_type string, nationality string, num_kill int);
truncate table gtd_master1;
insert into gtd_master1 (event_id, year, month, day, country, region, doubt_terrorist, success, suicide, attack_type, target_type, nationality, num_kill)
select event_id, year, month, day, country, (case when region='Southeast Asia' Then 'South Asia'
    when region = 'Australasia & Oceania' Then 'Oceania'
    when region = 'Central America & Caribbean' Then 'Latin America & the Caribbean'
    when region='South America' Then 'Latin America & the Caribbean'
    when region='Central Asia' Then 'Caucasus & Central Asia'
    when country = 'Mexico' Then 'Latin America & the Caribbean' else region end) as region, doubt_terrorist, success,
    suicide, attack_type, target_type, nationality, num_kill from gtd_raw;
```

Code Challenge: Getting aggregate results in GTD

We did not have any demographic data to control for number of events in relation to the population or the size of the country, so we decided to take percentages of some other factors such as successful events, events including suicide, based on the total number of events.

```
drop table if exists agg_gtd_master;
create external table agg_gtd_master (year int, country string, region string, num_events int, sum_mismatch int,sum_success int,
    sum_doubt int, sum_suicide int, sum_num_kill int, perc_suicide float, perc_doubt float, perc_success float, perc_mismatch float, killed_per_event float);
truncate table agg_gtd_master;
insert into agg_gtd_master select year, country, region, cast(count(event_id) as int), cast(sum(mismatch) as int), cast(sum(success) as int),
    cast(sum(doubt_terrorist) as int), cast(sum(suicide) as int), cast(sum(num_kill) as int), cast(((cast(sum(suicide)as int)/(cast(count(event_id) as int)))) as float),
    cast(((cast(sum(mismatch)as int)/(cast(count(event_id) as int)))) as float),
    cast(((cast(sum(mismatch)as int)/(cast(count(event_id) as int)))) as float),
    cast(((cast(sum(num_kill)as int)/(cast(count(event_id) as int)))) as float)
from gtd_agg_group by year, country, region;
```

Insights

Insights: General Insights

We ran correlations on different levels

- Year, country pairs available in both data sets 834 unique inputs.
- Countries across all the available years 139 unique inputs.
- Regions across all the countries and years 10 unique inputs.

Across all the correlations we ran, we found that the highest scores were obtained when exploring on the regional level. However, there are some outliers, which we will be present.

Insights: General Insights

We found very weak correlation of the general indices number of events, number of suicides, number of successful events with the general indices of Human Freedom score, Personal Freedom score and Economic Freedom score. This correlations was done on year, country pairs.

	num_events	sum_num_killed
hf_score	-0.259	-0.282
pf_score	-0.298	-0.318
ef_score	-0.149	-0.175

It seems that although week, the correlations are the strongest with the Personal Freedom score.

On year, country pairs there is **very minimal relationship between number of events and the scores for freedom of expression** - Expression & Information (pf_expression), Laws and Regulations that Influence Media Content (pf_expression_influence), Press Killed (pf_expression_killed), Press Jailed (pf_expression_jailed), Political Pressures and Controls on Media Content (pf_expression_control), State Control over Internet Access (pf_expression_internet). The **strongest (negative) correlation is with the score for Press Killed**.

	pf_expression	pf_expression_i nfluence	pf_expression _killed	pf_expression _control	pf_expression _internet
num_events	-0.151	-0.088	-0.297	-0.183	-0.063

country		year	num_events	pf_expression		pf_expression_influence
Iraq	Ī	2016	3383	5.300000190734863	Ī	2.299999952316284
Iraq	İ	2017	2503	5.400000095367432	İ	2.299999952316284
Pakistan	İ	2013	2214	7.099999904632568	ĺ	3.700000047683716
Pakistan	İ	2014	2151	7.099999904632568	İ	3.700000047683716
Pakistan	İ	2012	1654	7.099999904632568	Ì	3.700000047683716
Pakistan	Ĺ	2015	1243	7.300000190734863	İ	3.700000047683716
India	İ	2016	1026	8	İ	6.300000190734863
Pakistan	İ	2011	1012	7.099999904632568	İ	4
India	ĺ	2017	967	8	İ	6.300000190734863
Ukraine	İ	2014	895	6.599999904632568	i	5.300000190734863

On the country level the correlation increases, however, it is still quite weak.

	pf_express ion	pf_expression_influe nce	pf_expression_ killed	pf_expression_ jailed	pf_expression_ control	pf_expression_ internet
num_events	-0.198	-0.130	-0.542	-0.004	-0.204	-0.099

country	num_events	pf_expression_c pf_expression_influence_c
Iraq	21590	5.350000381469727 2.299999952316284
Pakistan	11805	7.550000190734863 4.360000133514404
India	7559	8.079999923706055 6.630000114440918
Philippines	4403	8.140000343322754 5.789999961853027
Nigeria	3691	8.239999771118164 4.820000171661377
Yemen	3191	5.800000190734863 1
Thailand	2894	6.840000152587891 2.599999904632568
Libya	2236	5.860000133514404 3.099999904632568
Syria	2058	3.950000047683716 0.300000011920929
Egypt	1995	6.630000114440918 2.329999923706055

On the region level the **correlations increase**. The biggest increase is seen in the State control of the Internet. Given the steep jump, we are assuming that here the correlation is not a result of causation but there are other factors influencing the scores that we are just not aware of.

	pf_expression	pf_expression_ influence	pf_expression_ killed	pf_expression_ jailed	pf_expression_ control	pf_expression_ internet
num_events	-0.578	-0.561	-0.713	-0.565	-0.468	-0.674

region	num_events	pf_expression_r	pf_expression_influence_r
Middle East & North Africa South Asia Sub-Saharan Africa Eastern Europe Latin America & the Caribbean North America Caucasus & Central Asia East Asia	35128 29234 7944 2999 2144 1805 341 144 143	6.373831748962402 7.371716976165771 7.386792659759521 8.459375381469727 9.376859664916992 8.089691162109375 9.447368621826172 6.491666793823242 7.516666412353516 9.445454597473145	2.652336359024048 4.011111259460449 3.832075595855713 5.667187690734863 8.031405448913574 5.061855792999268 8.315789222717285 2.629166603088379 4.545833587646484 8.300000190734863

One of the **highest correlations we observed for year, country pairs** was for the correlations with the score for **Disapperances, Conflict & Terrorism** (**pf_ss_disappearances**). Correlation is even higher on the country and region level. These are all **negatively correlated** - higher the number of the events, lower the score.

	pf_ss	pf_ss_women	pf_ss_disappearances
num_events	-0.338	-0.186	-0.530

pf_ss is the Security and Safety score and pf_ss_women is the SS score for women. Both pf_ss_womena and pf_ss_dissapearences are aggregated with other scores in the pf_ss.

Table showing top 5 year, country pairs ordered by the highest number of events.

	year	country	region	1	num_events	pf_ss_disappearances
+ -	+		+	-+		+
1	2016	Iraq	Middle East & North Africa	1	3383	0
	2017 j	Iraq	Middle East & North Africa	İ	2503	0
	2013 j	Pakistan	South Asia	İ	2214	3.5
	2014	Pakistan	South Asia	İ	2151	3.799999952316284
	2012 j	Pakistan	South Asia	i	1654	3.5

Table showing bottom 5 year, country pairs ordered by the highest number of events.

region	. –	pf_ss_disappearances
Latin America & the Caribbean	1	9
Western Europe	1	i 10
Sub-Saharan Africa	1	9
Latin America & the Caribbean	1	9.5
Caucasus & Central Asia	1	9
	Latin America & the Caribbean Western Europe Sub-Saharan Africa Latin America & the Caribbean	Latin America & the Caribbean 1 Western Europe 1 Sub-Saharan Africa 1 Latin America & the Caribbean 1

Correlation on level of countries (across all the available years). This is an increase from that on year, country pairs.

	pf_ss	pf_ss_women	pf_ss_disapperances
num_events	-0.359	-0.195	-0.623

country	region	. –	pf_ss_disappearances_avg_c
Iraq	Middle East & North Africa		0
Pakistan	South Asia	11805	4.190000057220459
India	South Asia	7559	7.090000152587891
Philippines	South Asia	4403	6.679999828338623
Nigeria	Sub-Saharan Africa	3691	5.840000152587891

Correlation on level of regions (across all the available years and countries). This is an increase from that on the country level or year, country pairs.

	pf_ss	pf_ss_women	pf_ss_disapperances
num_events	-0.557	-0.738	-0.825

region		num_events	pf_ss_disappearances_avg	
	+		+	
Middle East & North Africa	1	35128	6.558677673339844	
South Asia	Ì	29234	7.584465980529785	
Sub-Saharan Africa	Ĺ	7944	7.617391109466553	
Eastern Europe	i	2999	8.971621513366699	
Western Europe	i	2144	9.735537528991699	

Insights #3: Terrorism is rather harder to define, and not necessarily determined in relationship to the losses of lives

Currently, there is no universal, world-wide agreed upon definition for terrorism.

Global Terrorism Database (GTD), which defines terrorism as "acts of violence by non-state actors, perpetrated against civilian populations, intended to cause fear, in order to achieve a political objective"

Based on our analytics of doubts_terrorist and success, this concept of "fear" is not shown to be directly influenced by losses in human lives, ie. fatalities occurred or if the outcome has resulted into death of the actors or the victims.

Insights #3: Terrorism is rather harder to define, and not necessarily determined in relationship to the losses of lives

- High correlation with doubt in the event was a terrorist attack and whether the event was successful, number of fatalities and whether it had led to suicide
- High correlation with success in a terrorist attack and whether the event was doubted
 in it being a terrorist attack, number of fatalities and whether it had led to suicide
- Both doubts in the terrorist attack and whether an event was successful or not were not heavily but still moderately correlated to the attacker being a non-national

	success	num_killed	suicide	Non_national terrorist
doubt_terrorist	0.857	0.750	0.716	0.540
success		0.886	0.847	0.443

Insights #4: There is minimal correlation between events of terrorism and freedom of religion

	pf_religion	pf_religion_estop	pf_religion_harassment	pf_religion_restrictions
num_events	-0.172	-0.095	-0.050	-0.086
sum_num_kill	-0.158	-0.048	-0.091	-0.042

Pf_religion in the target type of Private Citizens & Property, Business Government (General), Government (Diplomatic), Military, Religious Figures /Institutions

	Private Citizens & Property	Business	Government (General)	Government (Diplomatic)	Military	Religious Figures /Institutions
num_events	-0.125	-0.082	0.001	0.115	-0.168	-0.085

Obstacles

Obstacles #1: Missing Data

HFI Dataset has missing data, that was simply denoted by '-'. To get around that we replaced all the '-' with -1 in the cleaning stage and made sure to control for that in our analytic.

Obstacles #2: Non-compatible Data Sets

As described in the Data Challenges our data sets were not compatible. The names of the countries and regions were different. Some countries were event classified in different regions between the two datasets (Greece, Cyprus and Mexico). Furthermore, GTD had a list of events, where as HFI has aggregate results by year and country.

Hence, we had to aggregate results in GTD by year, country combinations and only keep those *year*, *country* combinations that were present in both. This ended up being **834 combinations across 10 regions and years 2008 - 2017, with 139 distinct countries** (average 6 years of data per country).

Given how varied these data points were, some countries had more years of data than others, it definitely impacted our results. Hence, we found it hard to draw any conclusive outcomes.

Obstacles #2: Non-compatible Data Sets

Example how varied the number of years per country is.

2008	Greece	Western Europe	53	i
2009	Greece	Western Europe	115	i
2010	Greece	Western Europe	49	i
2011	Greece	Western Europe	11	İ
2012	Greece	Western Europe	23	İ
2013	Greece	Western Europe	54	İ
2014	Greece	Western Europe	26	İ
2015	Greece	Western Europe	31	İ
2016	Greece	Western Europe	30	İ
2017	Greece	Western Europe	44	ĺ
2012	Guatemala	Latin America & the Caribbean	1	Ì
2013	Guatemala	Latin America & the Caribbean	6	Ì
2013	Guinea	Sub-Saharan Africa	2	ĺ
2015	Guinea	Sub-Saharan Africa	2	İ
2016	Guinea	Sub-Saharan Africa	1	ĺ
2009	Guinea-Bissau	Sub-Saharan Africa	1	ĺ
2012	Guinea-Bissau	Sub-Saharan Africa	1	1
2008	Guyana	Latin America & the Caribbean	3	1
2016	Guyana	Latin America & the Caribbean	1	1
2016	Haiti	Latin America & the Caribbean	1	İ

Obstacles #2: Non-compatible Data Sets

Name changes that had to be made to HFI to match GTD notation:

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Yemen, Rep. (HFI) → Yemen (GTD)
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Slovak Rep. (HFI) → Slovak Republic (GTD)

Pap. New Guinea (HFI) → Papua New Guinea (GTD)

Kyrgyz Republic (HFI)→ Kyrgyzstan (GTD)

Timor-Leste (HFI) \rightarrow East Timor (GTD)

Korea, South (HFI) \rightarrow South Korea (GTD)

Bosnia and Herzegovina (HFI)→ Bosnia-Herzegovina (GTD)

Dominican Rep. (HFI) → Dominican Republic (GTD)

Central Afr. Rep.(HFI) → Central African Republic (GTD)

Czech Rep.(HFI) \rightarrow Czech Republic (GTD)

Congo, Rep. Of (HFI) \rightarrow Republic of the Congo (GTD)

Congo, Dem. R. (HFI) → Democratic Republic of the Congo (GTD)

Regions that had to be combined in the GTD to match HFI:

South Asia (HFI) → South Asia (GTD) + Southeast Asia (GTD)

Oceania (HFI) → Australasia & Oceania (GTD)

Latin America & the Caribbean (HFI) → Central America & Caribbean (GTD) +

South America (GTD)

Caucasus & Central Asia (HFI) → Central Asia (GTD)

<u>Countries with mismatched regions:</u>

Cyprus: Eastern Europe (HFI); Western Europe (GTD)

→ Western Europe

Greece: Eastern Europe (HFI); Western Europe (GTD)

→ Western Europe

Mexico: Latin America & the Caribbean (HFI); North America (GTD)

→ Latin America & the Caribbean

Obstacles #3: Lacking Research Question and Poor Planning

We were not able to find a lot of research that would look into the relations of terrorism and freedom, besides the transnational terrorism, hence we decided to focus on on getting wide array of different results instead of setting a specific research question. In hindsight, this was not the best approach, as we now have a lot of general insights but very little of those that have more significant impact.

Furthermore, this was a project like this we have both undertaken. We underestimated the importance of proper planning and data profiling. Had we done that part better, we would likely have more insightful results.

Obstacles #4: Conclusions

The most conclusive results ended up being those that were aggregated by region (10 in total) and not by year, country combinations. However, there are other factors that could influence that. Hence our results, while insightful, are hardly conclusive. Other factors, that we did not investigate but could have an effect on this are cultural, political, geographical, etc.

Furthermore, a lot of our conclusions are based on the correlations between number of terrorist events and freedom scores. Since we **do not control the number of events for the population** of the country/region or the number of the countries in the region, those are **likely skewed towards the regions with higher population**.

Reflection

After the completing the analytics part of the project, we realized there were much room for improvement. Below are some of the things **we would rethink**, were we starting this project all over again.

- We would **define the research question better**. That would allow us to focus this project more narrowly and go in depth instead of breadth.
- We would **set up the experiments better** and and find a way to compare the correlations as well by having them structured in a separate table.
- We would explore the use of other tools, besides impala, such a tableau or similar that would allow us to do some basic visualizations as well.
- We would add demographic data as an additional dataset. That would enable us to do better and more in depth correlations measures.
- We would do a **case study**. Exploring these correlations for a particular country or region.

Summary

- Broad overview of the correlations between the instances of terrorism accross the world with different freedom indices.
- Due to the limited scope of the project, we cannot make any definite conclusions on the causations, but the explored correlations are very insightful.
- No significant correlation between freedom of press and number of terrorist attacks.
- No significant correlation between freedom of religion and number of terrorist attacks.

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