analysis-projectv2

January 16, 2025

0.1 Analysis of Battle-related deaths during natural resource conflicts in Latin-America from 1989 to 2006

0.2 1. Introduction

Nowadays, wars are still occurring, with the main reason for invasion being natural resources. Unfortunately, the locals are also affected, which leads to an increasing number of deaths. The aim of this project is to distinguish the most devastating years in which the greatest number of civilians were killed, as well as the countries in Latin America where the most calamitous wars happened from 1989 to 2006. The study also helps to identify the main causes of protracted conflicts over natural resources. The area of research is a Latin America, which is generally understood to include the entire continent of South America, plus Mexico, Central America and the islands of the Caribbean.

0.3 1. Questions

- 1. Which were the top 5 most violent opposing groups, who killed the most number of civilians?
- 2.1. What were the top 5 years/countries in which most civilians were killed in natural resource conflicts in Latin America?
- 2.2 What were the mechanisms by which violent natural resource conflicts were affected from 1989 to 2006?

0.4 2. Data Sources

0.4.1 2.1 Descriptions of Data Sources

Datasource Name	Description	Avail	Data abi Ttyp e	Geographic Coverage
The Natural Resource Conflict Dataset	Codes whether internal armed conflicts are clearly linked to natural resources from 1946 - 2006.	[1]	Stata	Global, covering multiple regions

Datasource Name	Description	Availa	Data bi Tiyy pe	Geographic Coverage
UCDP Battle-Related Deaths Dataset version 24.1	Contains data on battle deaths (soldiers and civilians killed in combat) in state-based conflicts for 1946–2008.	[2]	Zipped CSV	Global, covering multiple regions

0.4.2 2.2 Licenses and Permissions

Since both datasets, The Natural Resource Conflict Dataset and UCDP Battle-Related Deaths Dataset version 24.1, belong to The Peace Research Institute Oslo (PRIO), we could look at licences and permission denoted by this organization. At this web page you could find this information:

...open data policies benefit Norwegian social science - and PRIO in particular...All data collection efforts at PRIO are openly available... from [3]. This proves that they are under a standard open-data license. Additionally, it was included that, in order to use theirs datasets for a research purpose, you need to include the references to these papers, where they were first presented. You could cite it from [1] and [2].

0.4.3 2.3 Data Pipeline

ETL (Extract, Transform, Load) Process:

The data pipeline was implemented for two datasets using Python. It consists of several steps provided below:

Pipeline Name	Extract	Transform	Load
Natural	Reads metadata	Filters rows, removes unnecessary	Stores
Resource	and downloads	columns, standardizes data format, filters	transformed
Conflict	Stata files.	manually for Latin-America countries,	data in
Dataset		removes empty cells and duplicates, adds	SQLite
Pipeline		a years column from start_date and	database as
		end_date.	conflicts.sc
UCDP Battle-	Reads zipped	Filters rows, removes unnecessary	Stores
Related	metadata and	columns, filters automatically for	transformed
Deaths	downloads CSV	Latin-America countries, renames	data in
Dataset	files.	location_inc to location for	SQLite
Pipeline		consistency with the first pipeline,	database as
version 24.1		removes empty cells and duplicates.	deaths.sqlit

More information regarding which transformations and cleaning steps were done and why you could find in a previous section 2.2 Structure and Quality of Data Sources.

I encountered with the problems: - when I tried to filter my datasets for Latin-America countries without a region information, which made me to search for the region of the country manually using maps; - when I tried to standardize data format into 'years', 'days', 'months' and 'whole days' from data string type; - to decide which columns are meaningful for my task.

Regarding Handling Errors or Changing Input Data, I used: - prevent_errors function to remove all empty rows and duplicates using pandas methods: dropna and drop_duplicates - a function which removes rows where the end_date - start_date = 0or where the whole_days = 0 - I changed input data of the dataset 1, which is a datatype string, in order to achieve consistency with the dataset 2. Finally we got a new column year to check correspondence with each database.

0.4.4 2.4 Combined Table

Table 1. First row of final Dataset, which combines The Natural Resource Confl

				DistributiAggravati@inancing			
Opposing				Mech-	Mecha-	Mech-	
$location Governmen \hbox{\it Group}$	Year	Dea	th Whol Beginnys	End	anism	nism	anism
Guatem Glovernmen URNG of Guatemala	1989	82	10,95 7 996	1995	1	0	0

The final table was created after applying our ETL pipeline to two previously mentioned datasets. By using SQLite operations INNER JOIN, it was saved as a CSV file combined_dataset.csv.

0.4.5 2.5 Answering the questions

Using SQLite queries I get the results in the forms of tables, which were saved as CSV files, e.g. "quest{}.csv".

Fig 1. Top 5 most violent Opposing Groups
 In Figure 1, the bar chart shows the top 5 most violent opposing groups from 1989 to 2006.
 However, the most bloody aggressor forces were the ELN and Farc in Colombia, who killed 21586 people in 17 years.

2.

Fig 2. Top 5 most violent wars based on year and location

In Figure 2, the scatter plot shows the top 5 most violent wars by year and location. Thus, the worst situation was in 1989 in El Salvador, where the largest number of civilians killed was 4924.

3.

Fig 3. Mechanisms over Location and number of killed civilians

In Figure 3, the scatter plot shows the mechanisms by which violent natural resource conflicts were affected from 1989 to 2006. Thus, the worst conflicts were in Colombia, where the largest number of civilians killed over 17 years was 21586. It happened due to financing mechanism, where Rebel groups use natural resources to finance rebellion. The second most cruel wars happened in Peru (6482 killed) due to financing mechanism and aggravation mechanism, where natural resource issues aggravate an ongoing conflict. The Third most cruel natural conflict was in El Salvador (5861 killed) due to distribution mechanism, where disagreements arise over distribution of natural resources or revenues. In conclusion, the most common cause of natural conflicts, which leads to enourmous amount of deaths among civilians, was due to financing mechanism.

0.5 3. Results and Limitations

• Results:

- Datasets stored in SQLite include only Latin-America countries. They were stored in a that way due to necessity of combining 2 different tables, that could be done using foreign keys, e.g. year and location.
- The datasets are complete by removing empty and duplicates values
- They are consistent in terms of same columns of a year and a location across both datasets
- They have same timeliness in terms of the year column.

• Limitations:

- Timeliness issue: The dataset 2 [2] does not include the starting and ending dates, but rather years. For this reason, we could depend regarding precise days on the data given from the table 1 [1].

0.6 References

- 1. The Natural Resource Conflict Dataset- Web page.
- 2. UCDP Battle-Related Deaths Dataset version 24.1 Web page.
- 3. PRIO license.
- 4. UCDP Battle-Related Deaths Dataset version 24.1 Link to the article.