College of Computer and Information Sciences Department of Information Technology

IT 362 Course Project Semester-2, 1446H



# IT362: Principles of Data Science

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# Phase 2: Data Collection, Processing, Cleaning, and Exploratory Data Analysis (EDA)

(note: everything mentioned in the report was also explained in the jupyter notebook)

### Prepared by

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# 1 Primary Data

#### **Overview**

The primary dataset contains less than 100 entries with 15 columns, including categorical and numerical variables. Key columns include:

• Event Name: 20 non-null values (high missingness)

Country: 91 non-null values
 ISO Code: 90 non-null values

• **Disaster Group**: Fully populated

• Geographical Coordinates: Latitude and Longitude available for all entries

• Temporal Data: Start and End Year, Month, and Day

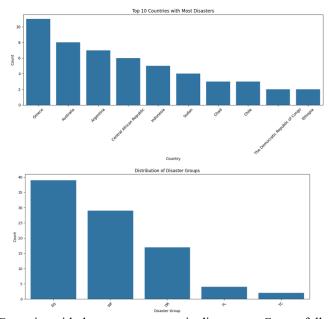
• Magnitude & Unit: Present in most cases

• Losses: Completely missing

#### **Key Observations**

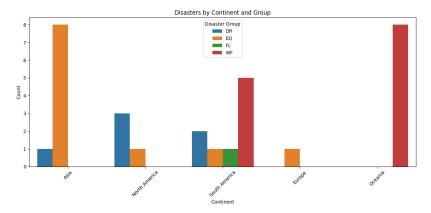
- There is substantial missing data in Event Name and Losses, potentially requiring imputation or removal.
- The dataset covers a diverse range of disaster events across multiple countries.
- Temporal data is complete, enabling trend analysis over time.
- Geographical data allows for spatial visualization of disasters.

#### **Visualizations**

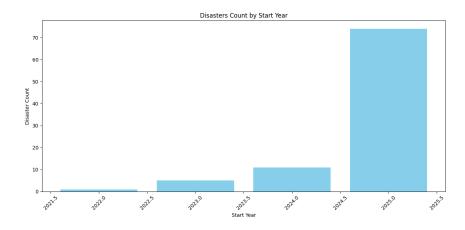


Countries with the most occurance in diasters are Greece followed by Australia and Argentina

From the dataset exploration, we observed that **EQ** (**Earthquake**) is the most frequent disaster type, followed by **WF** (**Wildfire**).



- Wildfires (WF) occur most frequently in Oceania, followed by South America indicating a significant climate-driven disaster pattern in the region.
- Earthquakes (EQ) are most prevalent in Asia, highlighting the region's vulnerability to seismic activity.
- Other disaster groups show less occurrences across continents, but these two stand out as the most frequent occurrences.



# 2 Secondary Data

#### **Overview**

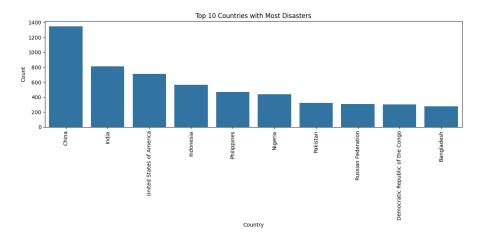
The secondary dataset provides pre-existing information that complements the primary dataset. This includes:

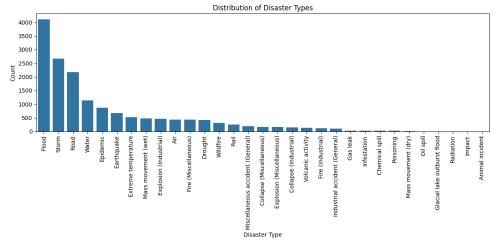
- Historical Disaster Records: Potentially overlapping with primary data
- Economic Impact Data: Used to infer losses where missing in the primary dataset
- Geographical Information: Enhancing location-based insights

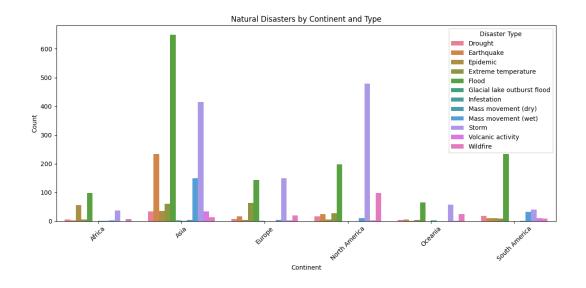
#### **Key Observations**

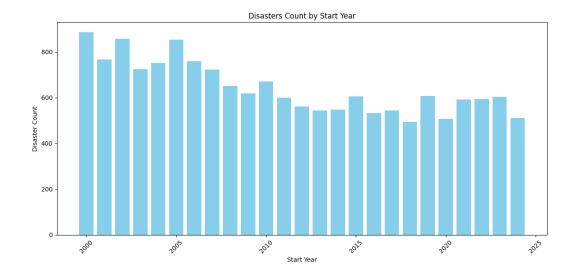
- Some discrepancies exist between recorded disasters in primary and secondary datasets.
- Losses information is more complete, allowing potential imputation for primary data.
- Differences in categorical classifications may require standardization.

#### **Visualizations**









Upon comparing the **primary dataset** with the **secondary dataset**, we observed a significant difference in the **number of rows**, **attributes**, **and overall structure**. To ensure a **comprehensive analysis**, we proceeded with **EDA on both datasets** to extract key insights and compare them effectively.

However, for a more **unified and enriched analysis**, we integrated both datasets into a single **combined dataset**. This allowed us to capture a broader perspective while maintaining consistency in our exploration. The EDA was then primarily focused on the **integrated dataset**, ensuring that all relevant patterns and trends were thoroughly examined.

# 3 Comparison: Primary vs. Secondary Data

#### **Alignments**

- Both datasets capture similar disaster events and geographical data.
- Temporal trends largely align, reinforcing the reliability of time-based analyses.

#### **Discrepancies**

- Missing values in the primary dataset contrast with more complete data in the secondary source.
- Differences in classification schemes for disaster types.
- Some inconsistencies in recorded magnitudes.

#### **Comparison**

Aspect	Primary Data (API)	Secondary Data (Historical)
Row Count	<100	16,000+
Disaster Count	5 types	Over 10 types
Time Period	2020 and later	1900s and later
<b>Most Affected Country</b>	Greece (10 disasters)	China (1,200+ disasters)
Most Frequent Disaster	Varies by country; Asia → Earthquakes  Oceania → Wildfires	Asia → <b>Floods</b> (most frequent overall)
Disaster Type Mapping	Limited to news-reported disasters	Covers a wider variety of disasters

# 4 Summary of New Insights and Hypotheses

#### **Insights**

- The primary dataset requires imputation, especially for financial losses.
- The secondary dataset can act as a validation tool for historical disaster trends.
- Discrepancies highlight the importance of standardization in disaster reporting.

#### **Hypotheses**

- The missing Losses column in the primary dataset can be estimated using secondary data trends.
- Different disaster classification schemes affect analytical conclusions and should be harmonized.
- Temporal and geographical patterns can help predict disaster occurrences and economic impacts.