

# Global Terrorism Insights

## Comprehensive Analysis of Global Terror Data

Capstone Project Presentation

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# Introduction

- Terrorism continues to be a **complex global challenge** requiring advanced analytical solutions
- This project leverages the **Global Terrorism Database (1970-2021)** with over 214,000 incidents
- The database contains **135 attributes** for each incident, providing rich data for analysis
- Our goal: Use **machine learning** to predict the terrorist group responsible for attacks
- Data-driven methods show potential to **enhance counter-terrorism strategies**

# Problem Statement

## Challenge

- ⚠ Many terrorist attacks remain **unattributed** due to absence of claims
- ⚠ This delays effective countermeasures and response planning

## Our Approach




- 💡 Build **machine learning models** to predict likely perpetrator groups based on attack characteristics
- 💡 Address challenges: Class imbalance, noisy data, and complex patterns

## Benefits




- ✅ Support **faster investigations**, better resource allocation, and more effective preventive strategies

# Dataset Overview

## Global Terrorism Database (GTD)

-  **Open-source database** covering terrorist events worldwide
-  Time Period: **1970 through 2021**
-  Scale: Over **200,000** documented incidents

## Key Information

-  Coverage: Both **domestic and transnational** terrorist incidents
-  Data includes: Date, location, weapons, target type, casualties, and responsible groups
-  Contains **135 attributes** for each incident

*Source: START (National Consortium for the Study of Terrorism and Responses to Terrorism)*

# Data Preprocessing

## Duplicate Detection

Checked for duplicate values in the dataset; **no duplicate records** were found

## Missing Values

Identified null values and retained columns with **at least 95%** data present

## Categorical Variables

Applied **One-Hot Encoding** for low cardinality (<30) and **Label Encoding** for high cardinality (>30)

## Feature Engineering

Extracted relevant features after removing columns with excessive null values

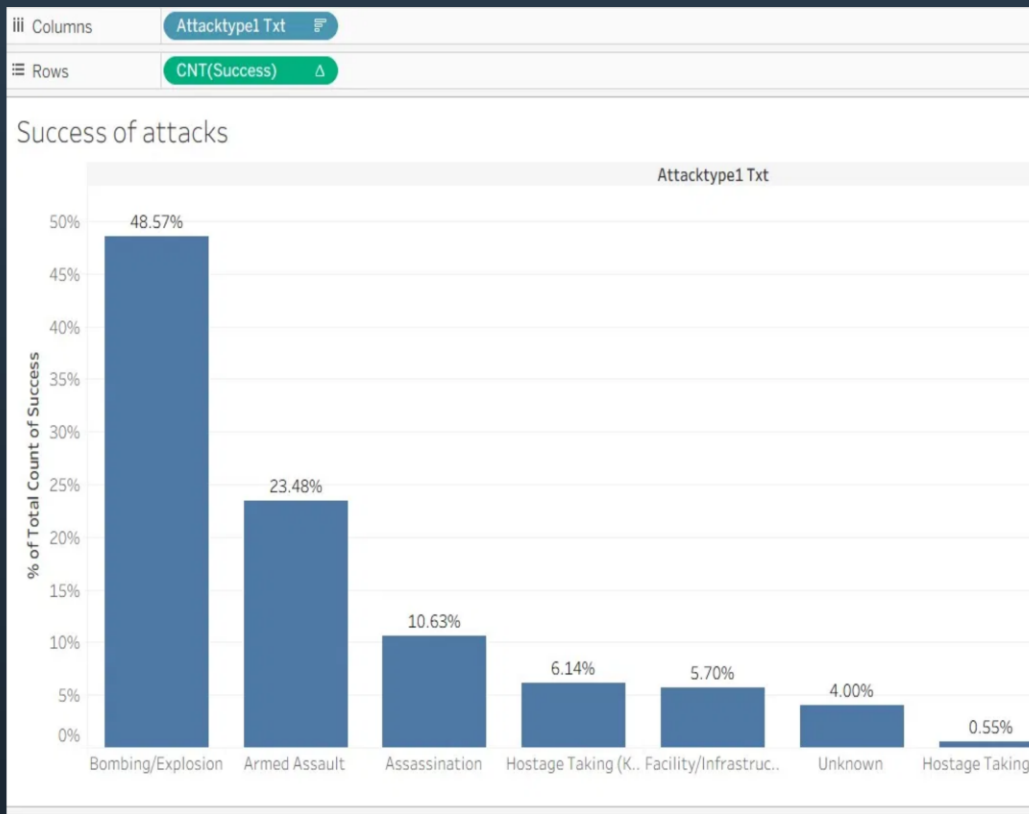
## Multicollinearity Treatment

Removed highly correlated columns to prevent redundancy

## Train-Test Split

Created separate datasets for **model training and evaluation**

# Attack Types Analysis



## Key Finding

**Bombing/Explosion** dominates with **48.57%** of successful attacks, followed by Armed Assault at **23.48%**

## Strategic Implications

Terrorists favor bombing tactics for their **high impact-to-effort ratio**, anonymity, and psychological effect

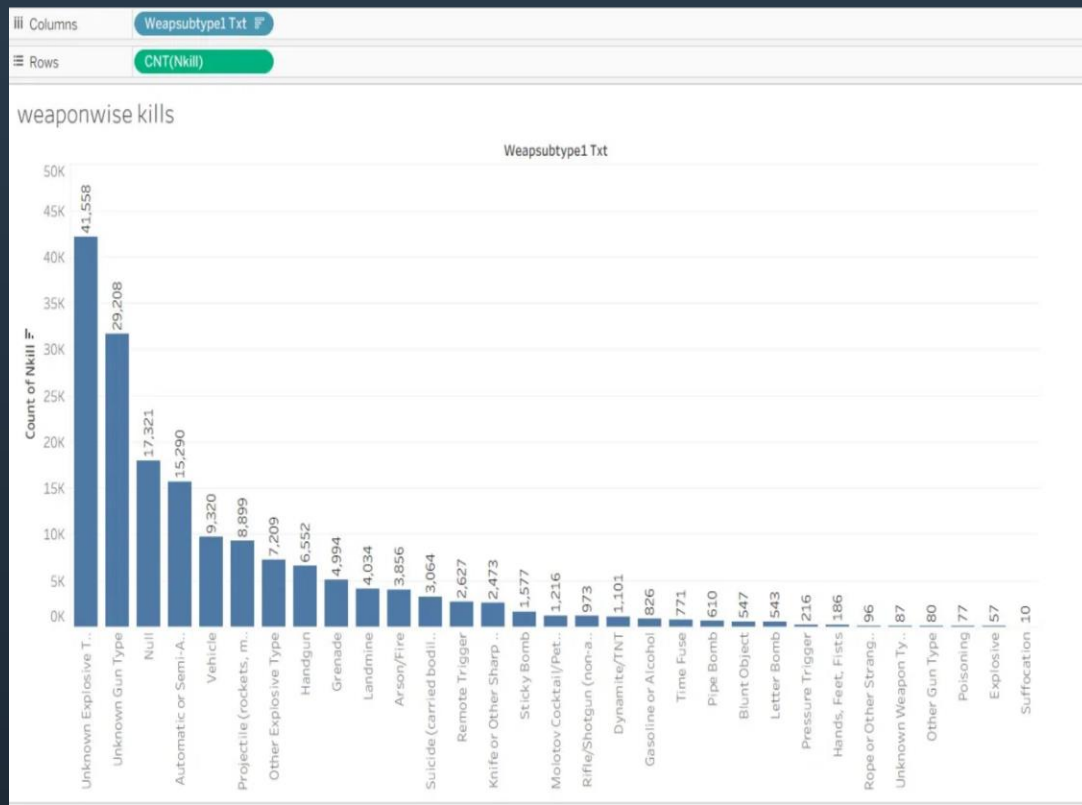
## Security Recommendations

Enhance **explosive detection** in public spaces and increase monitoring of **precursor materials**

## Trend Analysis

Low percentage of sophisticated attacks (e.g., hijacking at 0.36%) indicates preference for **high-impact, low-complexity methods**

# Weapons Analysis



## Key Finding

**Unknown Explosives (41,558)** and **Unknown Gun Types (29,208)** cause the most fatalities, highlighting significant intelligence gaps

## Weapon Preference Analysis

After unknown categories, **automatic weapons (17,321)** and **vehicle bombs (15,290)** are most lethal, indicating preference for high-casualty methods

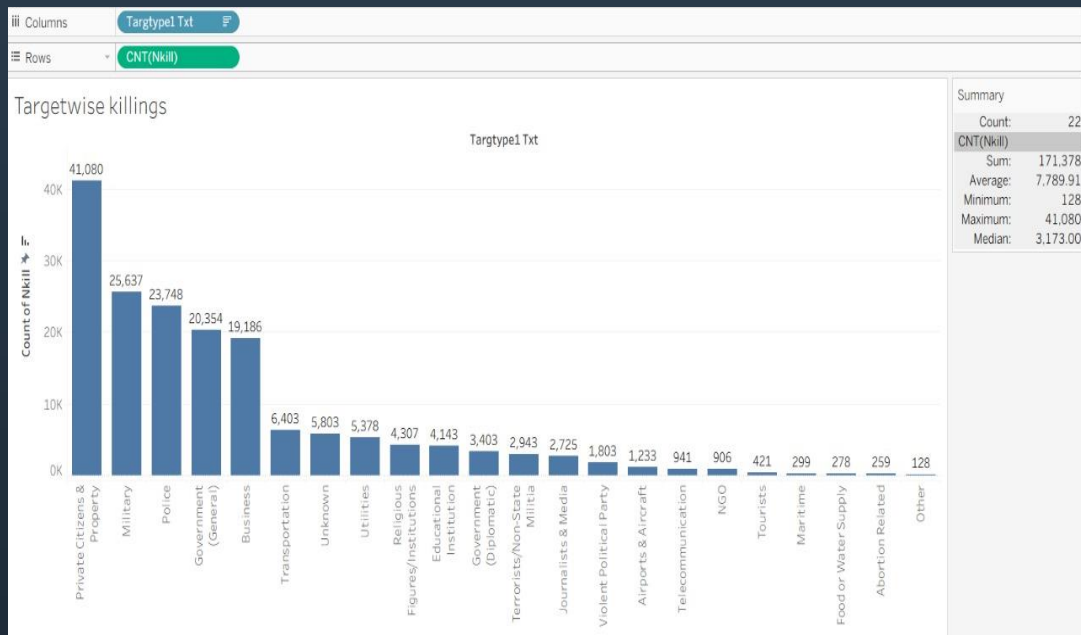
## Intelligence Implications

The high "unknown" categories suggest **forensic limitations** in post-attack investigations and need for improved weapon identification techniques

## Counter-Terrorism Strategy

Focus on **controlling access** to explosive materials and automatic weapons while

# Target Analysis



## Caption

This graph shows number of killings according to Target Type. From this we can infer that when target was private citizens & property the killings were higher. We see that least killings were in other target types that are not mentioned. Abortion related killings are the killings of doctors and medical staff who are involved in abortion related activities, by the protestors having ideology against abortion. Utilities killings indicates that attack on amenities like power house, dams, petrol pumps etc.

## Key Finding

**Private Citizens & Property** are overwhelmingly targeted with **41,080 fatalities**

## Target Hierarchy

Top five targets (Private Citizens, Military, Police, Government, Business) account for **76% of all fatalities**

## Strategic Pattern

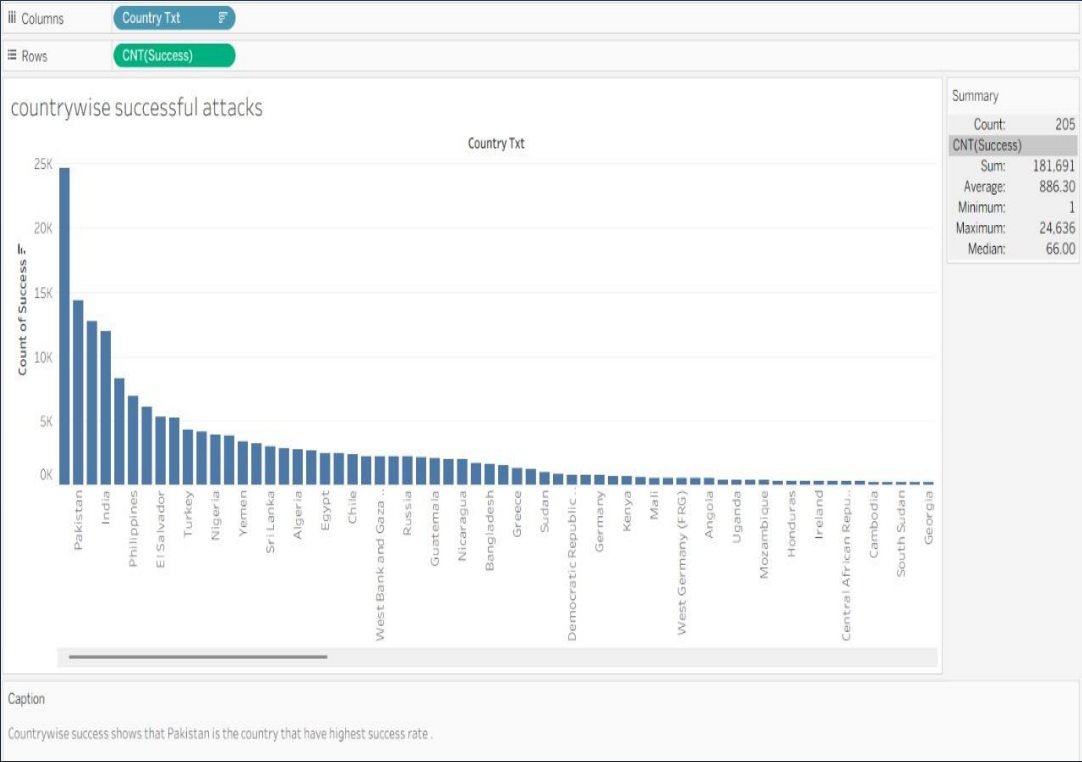
Targeting civilians creates **maximum psychological impact** while attacking authority symbols serves ideological goals

## Security Implications

Public spaces require enhanced **soft-target protection** strategies



# Geographic Analysis



## Key Finding

**Pakistan** leads with the highest number of successful attacks, followed by **Iraq**, **India**, and **Afghanistan**

## Regional Concentration

South Asia and Middle East regions account for the **majority of successful attacks**, indicating geopolitical hotspots South

## Success Rate Analysis

Afghanistan shows a particularly high success rate with **12,731 successful attacks**, suggesting operational advantages for terrorist groups in this region

## Security Implications

International counter-terrorism efforts should prioritize **regional cooperation** in these high-risk areas

# Model Evaluation



## Best Performer

**Random Forest** achieved highest accuracy (0.8167) and precision (0.8080), demonstrating superior classification ability.

## Ensemble Advantage

Ensemble methods (Random Forest, ADA Boost) consistently **outperformed** single-model approaches, highlighting the benefit of combining multiple decision trees.

## Performance Analysis







The 81.67% accuracy of Random Forest indicates strong but not perfect prediction capability, suggesting room for further model optimization.

## Application Implications

Models can serve as **investigative tools** to narrow down potential perpetrators, but should be used alongside traditional intelligence methods.

# Conclusion

## Key Findings & Implications

-  **Bombing/Explosion (48.57%)** is the dominant attack method, requiring specialized prevention strategies
-  **Pakistan, Iraq, and Afghanistan** emerge as critical regions requiring focused counter-terrorism efforts
-  **Random Forest** model achieved 81.67% accuracy in predicting terrorist group attribution
-  **Private citizens** are the most targeted group (41,080 fatalities), highlighting the need for public space security
-  High numbers of **unknown weapon categories** indicate critical intelligence gaps in forensic capabilities
-  Data science demonstrates significant potential for **improving counter-terrorism intelligence** and resource allocation

# Future Scope

## ⚙️ Advanced Model Optimization

Apply **hyperparameter tuning** and explore advanced ensemble techniques like xgboost to improve prediction accuracy

## 🗄️ Feature Enrichment

Integrate additional datasets such as **political stability indices**, economic indicators, and conflict databases

## 🖥️ Real-Time Prediction Systems

Deploy **cloud-based dashboards** that can take live incident data and predict likely perpetrators instantly

## 🧠 Deep Learning Applications

Experiment with **sequence models (LSTMs)** for temporal prediction and graph neural networks for analyzing terrorist organization connections

