

Women's Safety Abroad: A Country-by-Country Breakdown.

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Analyzing Women's Safety Across Countries

A Comprehensive Study on Women's Safety

Problem Statement

An overview of the critical issues surrounding women's safety globally.

Data Loading and Exploration

The initial phase of the project focusing on acquiring and exploring relevant data.

Data Cleaning

The process of ensuring the data is accurate and suitable for analysis.

Ourrelation Analysis

Analyzing relationships between various factors affecting women's safety. Composite Scores
Creation

Developing composite scores to quantify women's safety across different contexts.

Data Visualization
Techniques

Employing various visualization techniques to present the data effectively. Clustering
Methodology

Using clustering techniques to group similar data points for better insights. Recommendations for Preventive Measures

Offering actionable recommendations to enhance women's safety based on analysis.

Summary of KeyFindings

A concise overview of the main findings from the analysis conducted. Conclusion and Call

Wrapping up the presentation with a strong conclusion and a call to action.

Dataset Visualization



Exploration of the dataset is the foundational step to understand the structure, identify missing values, assess data quality, and discover initial insights.

Here is a visual representation of the dataset.

Tools used: Alpha Tably for Visualization, Python 3 (

Data Cleaning: Ensuring Data Integrity

Checking for missing Values and Dropping them if needed.

<class 'pandas.core.frame.DataFrame'> RangeIndex: 176 entries, 0 to 175 Data columns (total 11 columns): Column Non-Null Count Dtype 176 non-null object country MostDangerousCountriesForWomen WomenPeaceAndSecurityIndex Score 2023 176 non-null float64 MostDangerousCountriesForWomen WomensDangerIndexWDI TotalScore 2019 float64 50 non-null 50 non-null float64 MostDangerousCountriesForWomen WDIStreetSafety 2019 MostDangerousCountriesForWomen WDIIntentionalHomicide 2019 50 non-null float64 MostDangerousCountriesForWomen WDINonPartnerViolence 2019 50 non-null float64 MostDangerousCountriesForWomen WDIIntimatePartnerViolence 2019 50 non-null float64 MostDangerousCountriesForWomen WDILegalDiscrimination 2019 50 non-null float64 MostDangerousCountriesForWomen WDIGlobalGenderGap 2019 50 non-null float64 MostDangerousCountriesForWomen WDIGenderInequality 2019 float64 50 non-null 10 MostDangerousCountriesForWomen WDIAttitudesTowardViolence 2019 50 non-null float64 dtypes: float64(10), object(1) memory usage: 15.3+ KB

Basic descriptive statistics to get an overview of each numeric column.

Country	0
WPSI_2023	0
WDI_Total_2019	126
WDI_Street_Safety_2019	126
WDI_Intentional_Homicide_2019	126
WDI_NonPartner_Violence_2019	126
WDI_IntimatePartner_Violence_2019	126
WDI_Legal_Discrimination_2019	126
WDI_Global_Gender_Gap_2019	126
WDI_Gender_Inequality_2019	126
WDI_AttitudesTowardViolence_2019 dtype: int64	126

Correlation Analysis: Understanding Relationships

Significant Correlations

High Positive Correlations (Dangerous Factors):

- Legal Discrimination

 Gender Inequality

 Countries with discriminatory laws show higher gender inequality.

High Negative Correlations (Protective Factors):

- Street Safety
 Gender Inequality
 Societies with higher gender equality often report
 better street safety.

Peaceful societies score better on women's safety.



Correlation analysis helps identify relationships between variables, providing insights into how different factors contribute to women's safety or danger in various countries.

Composite Scores Creation: Quantifying Danger

Interpretation of Scores

Composite Danger Score: Highlights countries where violence, discrimination, and inequality are prevalent.

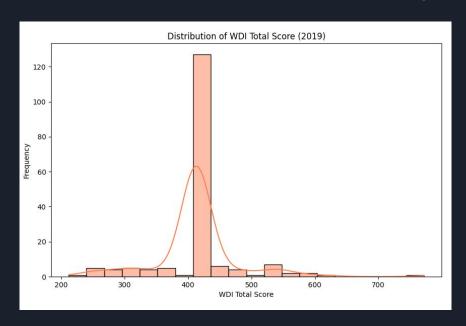
Composite Safety Score: Highlights countries with strong legal protections, societal safety, and equality.

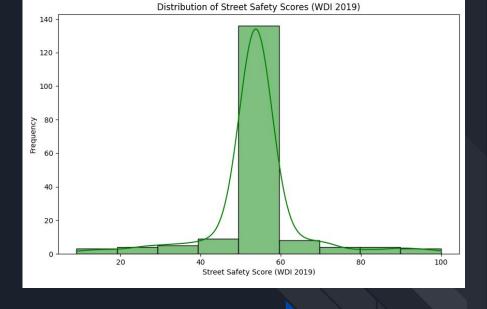
Why Use Composite Scores?

- 1. **Holistic Measurement**: Captures multiple dimensions of danger and safety.
- 2. **Actionable Insights**: Identifies areas requiring intervention.
- 3. **Simplification**: Provides a single measure for comparison across countries.

тор	10 countries l	by Composite Danger Score:
	Country	Composite_Danger_Score
23	South Africa	149.424444
6	Brazil	122.697778
8	Russia	115.412222
16	Iran	113.901111
39	Morocco	113.305556
12	Egypt	113.101111
10	Mexico	112.561111
0	India	112.361111
19	Thailand	112.024444
47	Saudi Arabia	108.928889
Тор	10 safest cour	ntries by Composite Danger Score:
	To surest cour	ici ico of composite banger score.
		Composite_Danger_Score
31	Country Spain	Composite_Danger_Score
31 121	Country Spain Ireland	Composite_Danger_Score 43.782222
31 121 96	Country Spain Ireland	Composite_Danger_Score 43.782222 50.812222
31 121 96 98	Country Spain Ireland Austria Switzerland	Composite_Danger_Score 43.782222 50.812222 52.292222
31 121 96 98 112	Country Spain Ireland Austria Switzerland	Composite_Danger_Score 43.782222 50.812222 52.292222 53.864444
31 121 96 98 112 116	Country Spain Ireland Austria Switzerland Singapore	Composite_Danger_Score 43.782222 50.812222 52.292222 53.864444 54.362222
31 121 96 98 112 116 90	Country Spain Ireland Austria Switzerland Singapore Norway Portugal	Composite_Danger_Score 43.782222 50.812222 52.292222 53.864444 54.362222 55.435556
31 121 96	Country Spain Ireland Austria Switzerland Singapore Norway Portugal	Composite_Danger_Score 43.782222 50.812222 52.292222 53.864444 54.362222 55.435556 55.620000

Data Visualization by Plotting using Python





Danger Score Distribution

- **Purpose**: Highlight countries with high danger scores.
- **Visualization:** Histogram with KDE (Kernel Density Estimate).

Street Safety Distribution

- Purpose: Understand the overall distribution of street safety scores across countries.
- Visualization: Histogram with KDE (Kernel Density Estimate).

Cluster Methodology

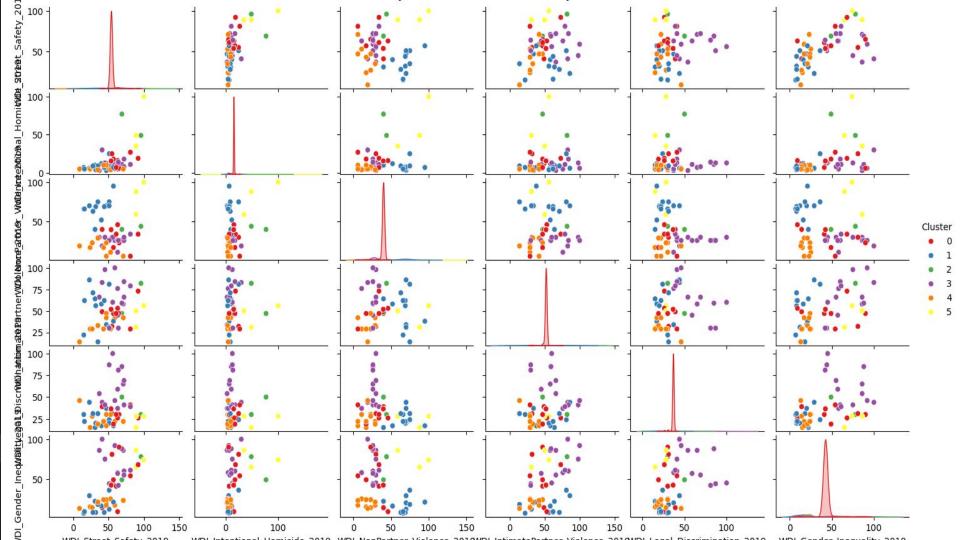
Clustering is an unsupervised machine learning technique that groups data points based on their similarities.

1. Grouping Countries by Risk Profile

- **Purpose**: Divide countries into clusters (e.g., low-risk, medium-risk, and high-risk).
- Benefits:
 - Simplifies complex data by summarizing it into categories.

2. Visualizing Relationships

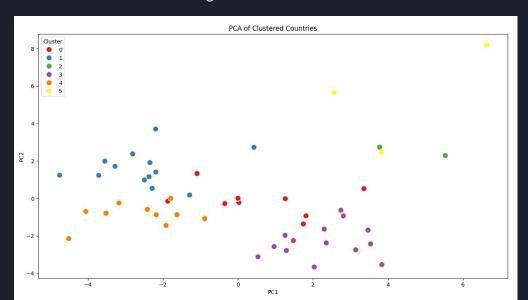
 Plotting clusters (e.g., using a scatter plot) makes it easier to understand how countries differ based on danger indicators like Street Safety, Legal Discrimination, and Gender Inequality.

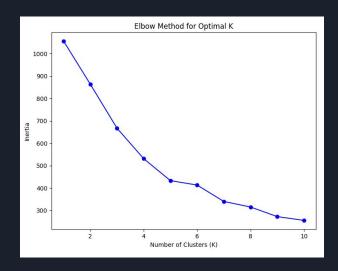


Elbow Curve and PCA of Cluster

How They Work Together

- 1. **Elbow Curve**: Determines the optimal number of clusters.
- 2. **PCA**: Reduces data dimensionality for:
 - Easier interpretation.
 - Effective visualization of the clusters in low-dimensional space.
 - Improved computational efficiency for clustering algorithms.





Together, these tools ensure that the clustering process is both statistically sound and visually interpretable, making it easier to derive actionable insights from the data.

Recommendations for Preventive Measures

1. Strengthen Legal Protections:

Enforce stricter laws against gender-based violence, including domestic violence and harassment.

2. Promote Gender Equality:

Reduce gender inequality through education, economic opportunities, and political representation.

3. Improve Street Safety:

Enhance public safety through better street lighting, CCTV surveillance, and increased police presence.

4. Raise Awareness and Education:

Conduct campaigns to challenge harmful societal norms and attitudes that perpetuate violence against women.

5. Strengthen Support Systems:

Train healthcare and law enforcement personnel to handle gender-based violence sensitively and effectively.

Summary and Conclusion

This project analyzes the factors contributing to the safety and danger levels for women in different countries, using indicators such as **Gender Inequality**, **Street Safety**, **Legal Discrimination**, and more. By clustering countries and ranking them based on composite danger and safety scores, the project identifies high-risk nations and provides data-driven insights.

Key findings:

- Countries with high danger scores often have poor gender equality and legal discrimination.
- Safety scores are strongly linked to effective governance and progressive societal norms.
- Clustering highlights distinct risk profiles, enabling tailored recommendations for each group.

Call to Action

For Governments:

- Prioritize reforms in legal protections for women.
- Allocate resources to enhance street safety and support systems.

For International Organizations:

- Provide funding and technical assistance to countries struggling with gender-based violence.
- Foster global awareness campaigns promoting women's safety and equality.

Conclusion

The analysis underscores the urgent need for legislative, societal, and infrastructure reforms to improve women's safety globally. High-risk countries require immediate attention, while medium- and low-risk nations should maintain and refine their policies.