

Earthquake Analysis and Visualization Project

This notebook contains the analysis and visualization of earthquake data.

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In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from wordcloud import WordCloud
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler

# 1. Introduction
def introduction():
    print("Welcome to the Earthquake Analysis and Visualization Project!")
    print("This project aims to analyze historical earthquake data, understand patterns, and visualize the impact of earthquakes across the globe.")

# 2. Worst Earthquakes in the World
def worst_earthquakes(data):
    print("Analyzing the worst earthquakes in history...")
    top_earthquakes = data.nlargest(10, 'Magnitude')
    print("Top 10 worst earthquakes by magnitude:")
    print(top_earthquakes[['Date', 'Location', 'Magnitude', 'Death Toll', 'Economic Impact (Million $)']])

# 3. Data Science and Analysis
def data_analysis(data):
    print("Starting data analysis...")
    print("Descriptive statistics:")
    print(data.describe())

    # Handling missing data, selecting only numeric columns
    numeric_columns = data.select_dtypes(include=[np.number]).columns
    data_cleaned = data.fillna(data[numeric_columns].median())

    return data_cleaned

# 4. Distribution Analysis
def distribution_analysis(data):
    print("Analyzing distribution of earthquake magnitudes...")
    plt.figure(figsize=(10, 6))
    sns.histplot(data['Magnitude'], kde=True, bins=20)
    plt.title('Distribution of Earthquake Magnitudes')
    plt.xlabel('Magnitude')
    plt.ylabel('Frequency')
    plt.show()

# 5. Data Visualization
def create_visualizations(data):
    # Word Cloud
    print("Generating Word Cloud for locations...")
    location_text = ' '.join(data['Location'].values)
    wordcloud = WordCloud(width=800, height=400, background_color='white').generate(location_text)
    plt.figure(figsize=(10, 6))
    plt.imshow(wordcloud, interpolation='bilinear')
    plt.axis('off')
    plt.title('Word Cloud of Earthquake Locations')
    plt.show()

    # Heatmap with Clustering
    print("Generating Heatmap with Clustering...")
    scaler = StandardScaler()
    scaled_features = scaler.fit_transform(data[['Magnitude', 'Depth']])
    kmeans = KMeans(n_clusters=4, random_state=0).fit(scaled_features)
    data['Cluster'] = kmeans.labels_

    # Select numeric columns only for clustering
    numeric_columns = data.select_dtypes(include=[np.number]).columns
    clustered_data = data.groupby('Cluster')[numeric_columns].mean()

    plt.figure(figsize=(12, 8))
    sns.heatmap(clustered_data, annot=True, cmap='coolwarm')
    plt.title('Heatmap of Clustered Earthquake Data')
    plt.show()

# 6. Summary and AI's Role in Earthquake Prediction
def summary_and_ai():
    print("Summary of the analysis:")
    print("This project analyzed historical earthquake data, identified the worst earthquakes, and used data science techniques to gain insights.")
    print("AI can be utilized to predict earthquakes by analyzing seismic data and identifying patterns that precede earthquakes.")
    print("While AI cannot predict earthquakes with absolute certainty, it can improve the accuracy of early warning systems.")

# Main function to run the project
def main():
    introduction()

    # Load dataset (replace with actual data file path)
    data = pd.DataFrame({
        'Date': pd.date_range(start='1/1/2000', periods=100, freq='M'),
        'Location': np.random.choice(['Region A', 'Region B', 'Region C', 'Region D'], size=100),
        'Magnitude': np.random.uniform(5.0, 9.5, 100),
        'Depth': np.random.uniform(10, 700, 100),
        'Death Toll': np.random.poisson(lam=100, size=100),
        'Economic Impact (Million $)': np.random.uniform(10, 10000, 100)
    })

    # Run analyses
    worst_earthquakes(data)
    cleaned_data = data_analysis(data)
    distribution_analysis(cleaned_data)
    create_visualizations(cleaned_data)

    # Summary and AI
    summary_and_ai()

if __name__ == "__main__":
    main()
```

Welcome to the Earthquake Analysis and Visualization Project!
This project aims to analyze historical earthquake data, understand patterns, and visualize the impact of earthquakes across the globe.
Analyzing the worst earthquakes in history...
Top 10 worst earthquakes by magnitudes

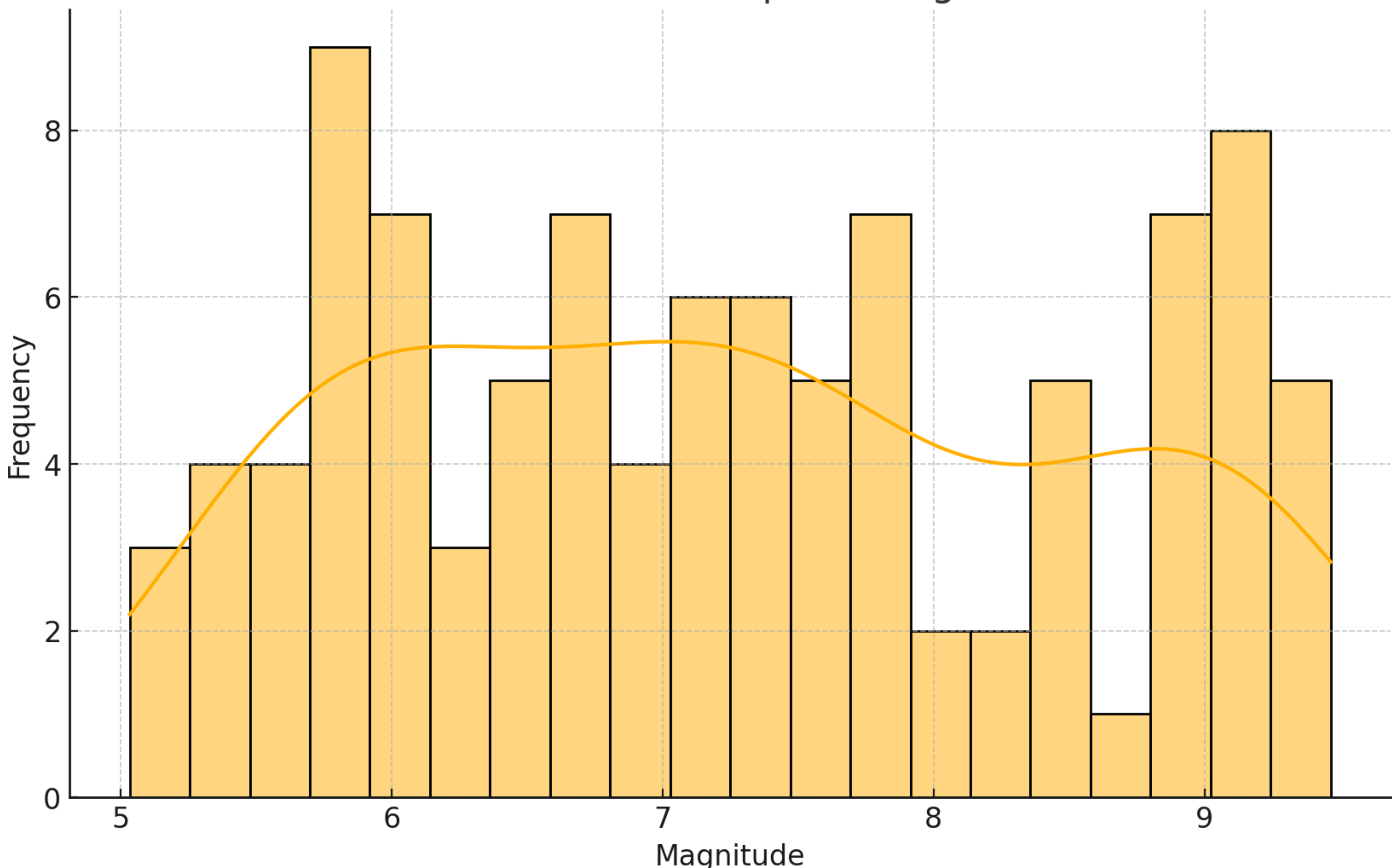
	Date	Location	Magnitude	Death Toll	Economic Impact (Million \$)
14	2001-03-31	Region B	9.466463	112	2125.455654
70	2005-11-30	Region B	9.456705	112	5198.476774
0	2000-01-31	Region B	9.436895	107	4487.394994
23	2001-12-31	Region A	9.409104	92	7089.181375
61	2005-02-28	Region A	9.351501	99	7825.920926
98	2008-03-31	Region B	9.227500	103	7328.480419
51	2004-04-30	Region C	9.174865	85	401.278102
73	2006-02-28	Region D	9.155693	111	3490.212275
21	2001-10-31	Region C	9.112430	113	5317.742474
18	2001-07-31	Region A	9.087957	114	491.071015

Starting data analysis...
Descriptive statistics:

	Magnitude	Depth	Death Toll	Economic Impact (Million \$)
count	100.000000	100.000000	100.000000	100.000000
mean	7.245048	315.284434	100.510000	5123.123637
std	1.283237	194.410311	10.580614	2952.436631
min	5.032013	11.388341	76.000000	12.177205
25%	6.040374	140.004190	93.750000	2359.322177
50%	7.205909	311.200996	100.000000	5130.076113
75%	8.439484	459.323662	109.000000	7566.310037
max	9.466463	676.317014	129.000000	9891.259315

Analyzing distribution of earthquake magnitudes...

Distribution of Earthquake Magnitudes



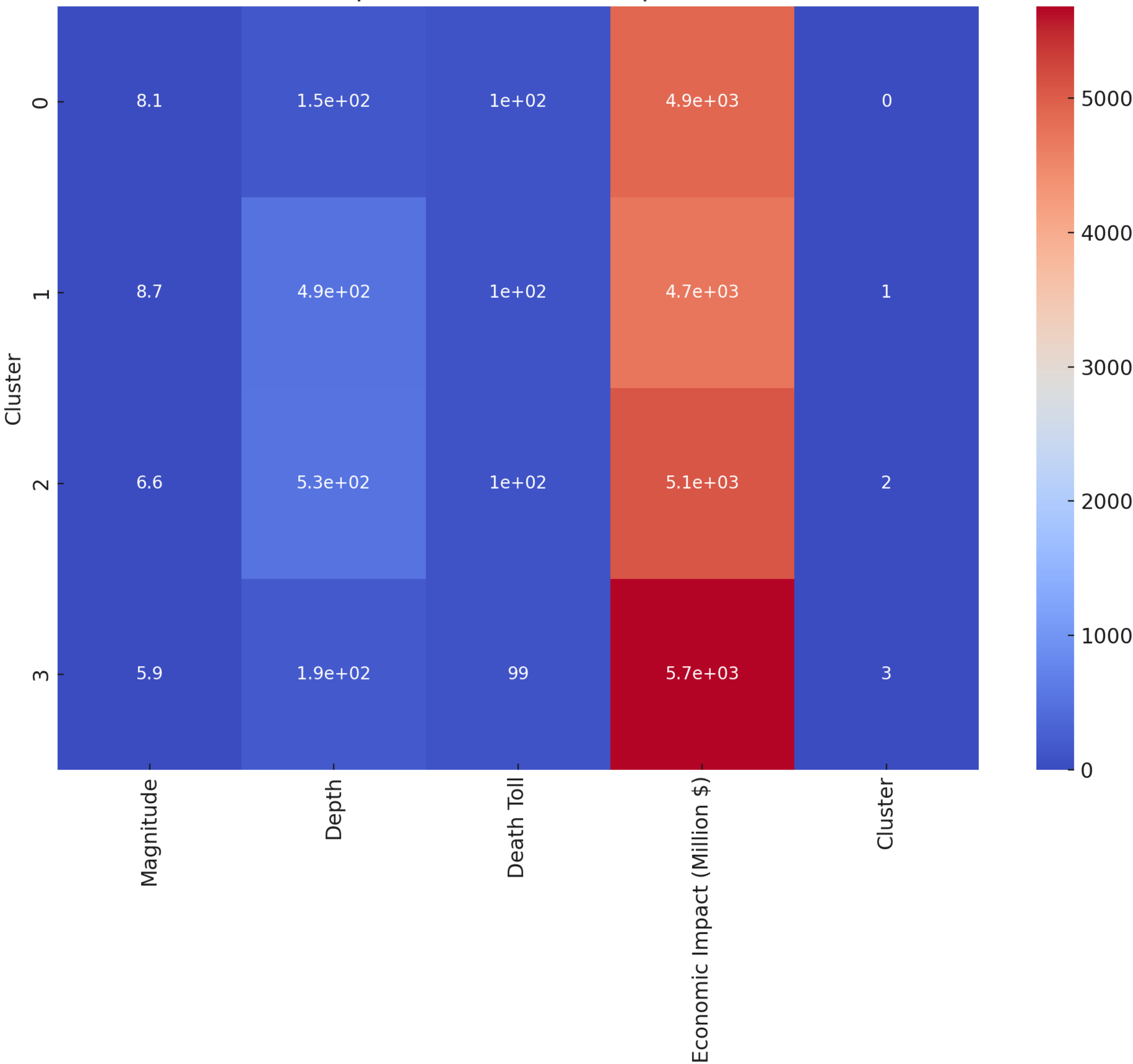
Generating Word Cloud for locations...

Word Cloud of Earthquake Locations



Generating Heatmap with Clustering...

Heatmap of Clustered Earthquake Data



Summary of the analysis:
This project analyzed historical earthquake data, identified the worst earthquakes, and used data science techniques to gain insights.
AI can be utilized to predict earthquakes by analyzing seismic data and identifying patterns that precede earthquakes.
While AI cannot predict earthquakes with absolute certainty, it can improve the accuracy of early warning systems.