

```
print("EARTHQUAKE MAGNITUDE DETECTION USING PRECURSOR SIGNALS")
print("DATASET - usa_earthquake_cleaned.csv")
```

EARTHQUAKE MAGNITUDE DETECTION USING PRECURSOR SIGNALS
DATASET - usa_earthquake_cleaned.csv

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import sqlite3
from datetime import datetime
import warnings
warnings.filterwarnings('ignore')

plt.style.use('default')
sns.set_palette("husl")

pd.set_option('display.max_columns', None)
pd.set_option('display.width', 1000)

print("Libraries imported successfully!")
print(f"Analysis Date: {datetime.now().strftime('%Y-%m-%d %H:%M:%S')}")
```

Libraries imported successfully!
Analysis Date: 2025-09-11 12:43:50

```
raw_data = pd.read_csv('../data/raw/usa_earthquake_data.csv')
cleaned_data =
pd.read_csv('../data/processed/usa_earthquake_cleaned.csv')
```

```
print("DATASET COMPARISON: ")
print(f"Raw Data Shape: {raw_data.shape}")
print(f"Cleaned Data Shape: {cleaned_data.shape}")
print(f"Columns Removed: {set(raw_data.columns) -
set(cleaned_data.columns)}")
print(f"Columns Added: {set(cleaned_data.columns) -
set(raw_data.columns)}")
```

```
print("\nCLEANED DATA SAMPLE: ")
cleaned_data.head()
```

DATASET COMPARISON:
Raw Data Shape: (9451, 22)
Cleaned Data Shape: (9451, 19)
Columns Removed: {'time', 'dmin', 'id', 'magNst', 'nst'}
Columns Added: {'time_only', 'date'}

CLEANED DATA SAMPLE:

	latitude	longitude	depth	mag	magType	gap	rms	net	
updated					place		type		
horizontalError	depthError	magError			status	locationSource			
magSource	date	time_only							
0	31.604000	-104.213000	4.4198	1.7	ml	69.00	0.50	tx	2024-01-26T05:08:27.774Z
					51 km NW of Toyah, Texas				
earthquake		0.00	1.292059		0.100	automatic			
tx	tx	1/26/2024	52:43.0						
1	64.501000	-146.905800	4.2000	1.4	ml	89.91	0.75	ak	2024-01-26T05:01:12.516Z
					2 km S of Salcha, Alaska				
earthquake		0.46	0.200000		0.173	automatic			
ak	ak	1/26/2024	42:50.7						
2	63.529000	-147.554300	13.1000	1.6	ml	89.91	0.62	ak	2024-01-26T04:34:54.160Z
					71 km ESE of Denali Park, Alaska				
earthquake		0.46	0.300000		0.173	automatic			
ak	ak	1/26/2024	32:51.5						
3	38.833168	-122.797165	1.7300	0.4	md	65.00	0.02	nc	2024-01-26T04:46:12.828Z
					6 km W of Cobb, CA				
earthquake		0.34	0.970000		0.310	automatic			
nc	nc	1/26/2024	29:01.2						
4	63.546200	-150.971900	0.0000	1.2	ml	89.91	0.80	ak	2024-01-26T04:25:43.831Z
					37 km E of Denali National Park, Alaska				
earthquake		0.46	0.400000		0.173	automatic			
ak	ak	1/26/2024	23:14.4						

```

print("\nDATA TYPES: ")
print(cleaned_data.dtypes)

print("\nMISSING VALUES: ")
missing_values = cleaned_data.isnull().sum()
missing_percentage = (missing_values / len(cleaned_data)) * 100
missing_df = pd.DataFrame({
    'Missing Count': missing_values,
    'Percentage': missing_percentage
}).sort_values('Missing Count', ascending=False)
print(missing_df[missing_df['Missing Count'] > 0])

print("\nNUMERICAL STATISTICS: ")
cleaned_data.describe()

```

```

DATA TYPES:
latitude          float64
longitude         float64
depth            float64
mag              float64
magType          object
gap              float64
rms              float64
net              object

```

```
updated          object
place            object
type             object
horizontalError  float64
depthError       float64
magError         float64
status           object
locationSource   object
magSource        object
date             object
time_only        object
dtype: object
```

MISSING VALUES:

Empty DataFrame

Columns: [Missing Count, Percentage]

Index: []

NUMERICAL STATISTICS:

	latitude	longitude	depth	mag	gap
rms	horizontalError	depthError	magError		
count	9451.000000	9451.000000	9451.000000	9451.000000	9451.000000
	9451.000000	9451.000000	9451.000000	9451.000000	
mean	39.946558	-110.580238	21.358118	1.631121	103.437011
	0.290898	1.628428	2.113351	0.226302	
std	18.996832	68.764250	50.809672	1.238299	54.598118
	0.281351	27.223453	9.483877	0.326050	
min	-65.436500	-179.968700	-3.170000	-1.270000	10.000000
	0.000000	0.000000	0.000000	0.000000	
25%	33.417833	-149.698400	2.816150	0.860000	70.000000
	0.100000	0.330000	0.400000	0.130565	
50%	38.817333	-122.447500	7.250000	1.400000	89.910000
	0.177100	0.460000	0.720000	0.173000	
75%	56.647900	-112.604333	14.300000	2.000000	119.000000
	0.460000	0.660000	1.500000	0.211872	
max	86.513000	179.996600	660.826000	7.500000	357.000000
	2.930000	2626.244209	494.700000	5.160000	

```
fig, axes = plt.subplots(2, 2, figsize=(16, 12))
```

```
scatter = axes[0,0].scatter(cleaned_data['longitude'],
```

```
cleaned_data['latitude'],
```

```
c=cleaned_data['mag'], cmap='Reds',
```

```
alpha=0.7, s=50)
```

```
axes[0,0].set_xlabel('Longitude')
```

```
axes[0,0].set_ylabel('Latitude')
```

```
axes[0,0].set_title('Earthquake Locations (Color = Magnitude)')
```

```
plt.colorbar(scatter, ax=axes[0,0], label='Magnitude')
```

```

axes[0,1].hist(cleaned_data['depth'], bins=25, alpha=0.7,
color='lightgreen', edgecolor='black')
axes[0,1].set_xlabel('Depth (km)')
axes[0,1].set_ylabel('Frequency')
axes[0,1].set_title('Distribution of Earthquake Depths')
axes[0,1].axvline(cleaned_data['depth'].mean(), color='red',
linestyle='--',
                    label=f'Mean: {cleaned_data["depth"].mean():.1f}
km')
axes[0,1].legend()

network_counts = cleaned_data['net'].value_counts().head(10)
axes[1,0].bar(network_counts.index, network_counts.values,
color='lightblue')
axes[1,0].set_xlabel('Seismic Network')
axes[1,0].set_ylabel('Number of Earthquakes')
axes[1,0].set_title('Top 10 Seismic Networks')
axes[1,0].tick_params(axis='x', rotation=45)

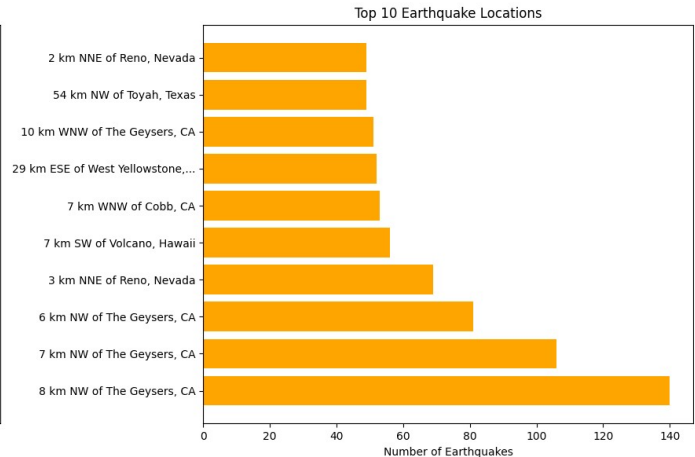
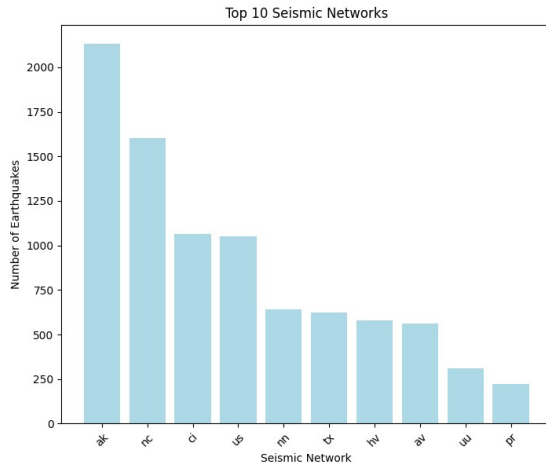
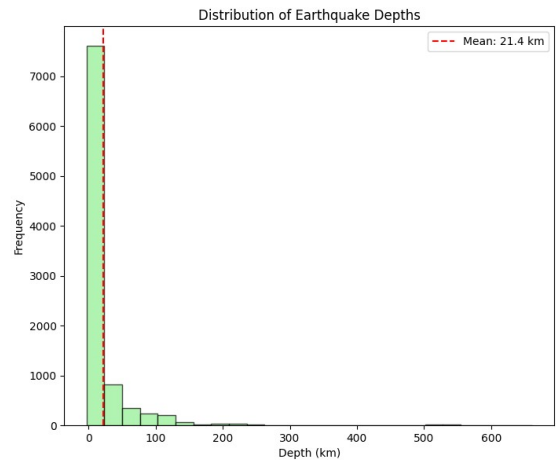
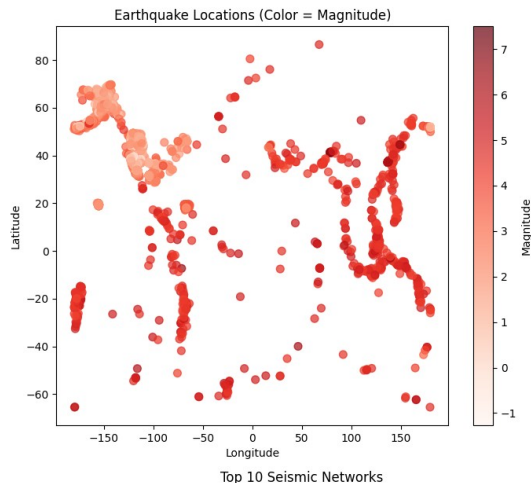
place_counts = cleaned_data['place'].value_counts().head(10)
axes[1,1].barh(range(len(place_counts)), place_counts.values,
color='orange')
axes[1,1].set_yticks(range(len(place_counts)))
axes[1,1].set_yticklabels([place[:30] + '...' if len(place) > 30 else
place for place in place_counts.index])
axes[1,1].set_xlabel('Number of Earthquakes')
axes[1,1].set_title('Top 10 Earthquake Locations')

plt.tight_layout()
plt.show()

print("GEOGRAPHIC STATISTICS: ")
print(f"Latitude Range: {cleaned_data['latitude'].min():.3f}° to
{cleaned_data['latitude'].max():.3f}°")
print(f"Longitude Range: {cleaned_data['longitude'].min():.3f}° to
{cleaned_data['longitude'].max():.3f}°")
print(f"Depth Range: {cleaned_data['depth'].min():.1f} -
{cleaned_data['depth'].max():.1f} km")
print(f"Average Depth: {cleaned_data['depth'].mean():.1f} km")

print(f"\nTop 5 Most Active Locations:")
print(cleaned_data['place'].value_counts().head())

```



GEOGRAPHIC STATISTICS:

Latitude Range: -65.436° to 86.513°
Longitude Range: -179.969° to 179.997°
Depth Range: -3.2 - 660.8 km
Average Depth: 21.4 km

Top 5 Most Active Locations:

Location	Count
8 km NW of The Geysers, CA	140
7 km NW of The Geysers, CA	106
6 km NW of The Geysers, CA	81
3 km NNE of Reno, Nevada	69
7 km SW of Volcano, Hawaii	56

Name: count, dtype: int64

```
numerical_cols = ['latitude', 'longitude', 'depth', 'mag', 'gap',
                  'rms', 'horizontalError', 'depthError', 'magError']
correlation_data = cleaned_data[numerical_cols].corr()
```

```
plt.figure(figsize=(12, 10))
mask = np.triu(np.ones_like(correlation_data, dtype=bool))
sns.heatmap(correlation_data, mask=mask, annot=True, cmap='coolwarm',
```

```

center=0,
        square=True, linewidths=0.5, cbar_kws={"shrink": 0.8})
plt.title('Correlation Matrix of Numerical Features')
plt.tight_layout()
plt.show()

fig, axes = plt.subplots(1, 3, figsize=(18, 5))

axes[0].scatter(cleaned_data['mag'], cleaned_data['magError'],
alpha=0.6, color='red')
axes[0].set_xlabel('Magnitude')
axes[0].set_ylabel('Magnitude Error')
axes[0].set_title('Magnitude Error vs Magnitude')

axes[1].scatter(cleaned_data['mag'], cleaned_data['horizontalError'],
alpha=0.6, color='blue')
axes[1].set_xlabel('Magnitude')
axes[1].set_ylabel('Horizontal Error (km)')
axes[1].set_title('Horizontal Error vs Magnitude')

axes[2].scatter(cleaned_data['depth'], cleaned_data['depthError'],
alpha=0.6, color='green')
axes[2].set_xlabel('Depth (km)')
axes[2].set_ylabel('Depth Error (km)')
axes[2].set_title('Depth Error vs Depth')

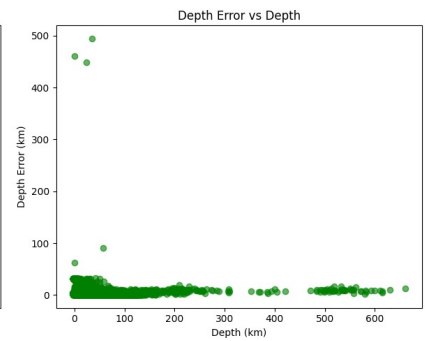
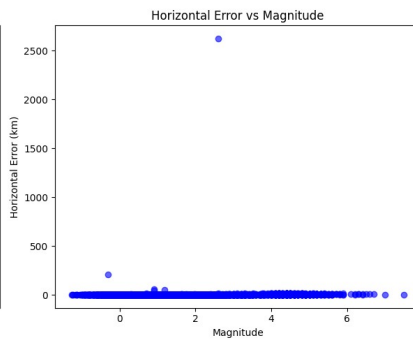
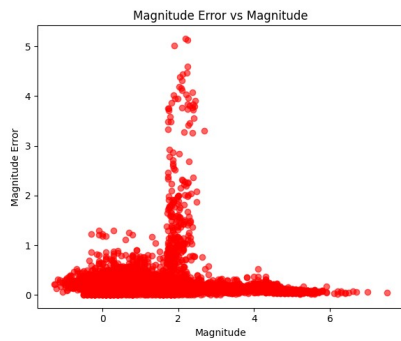
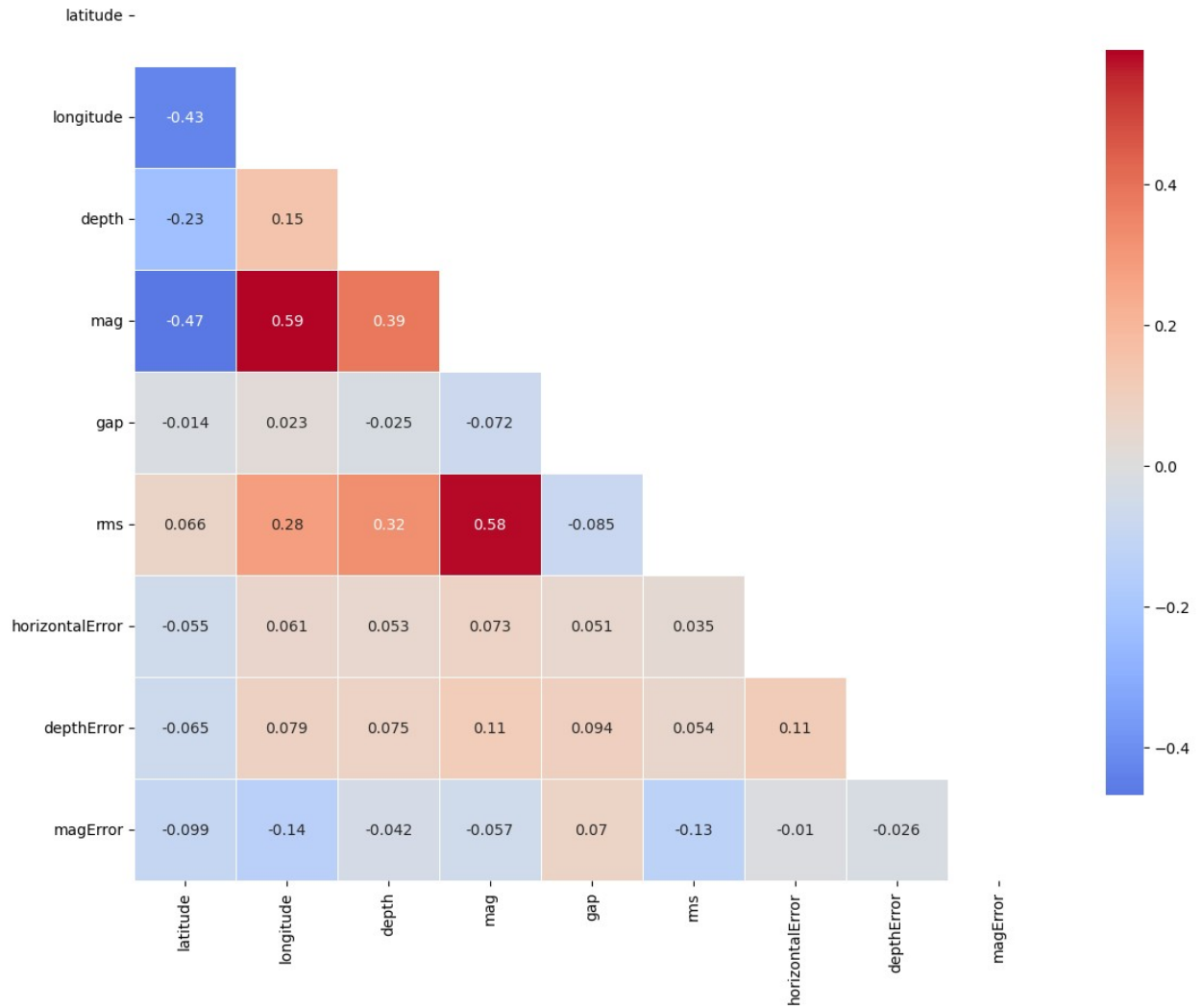
plt.tight_layout()
plt.show()

print("STRONG CORRELATIONS ( $|r| > 0.3$ ): ")
correlation_pairs = []
for i in range(len(correlation_data.columns)):
    for j in range(i+1, len(correlation_data.columns)):
        corr_val = correlation_data.iloc[i, j]
        if abs(corr_val) > 0.3:
            correlation_pairs.append((correlation_data.columns[i],
correlation_data.columns[j], corr_val))

for col1, col2, corr in sorted(correlation_pairs, key=lambda x:
abs(x[2]), reverse=True):
    print(f"{col1} - {col2}: {corr:.3f}")

```

Correlation Matrix of Numerical Features



STRONG CORRELATIONS ($|r| > 0.3$):

longitude - mag: 0.591

mag - rms: 0.582

latitude - mag: -0.469

latitude - longitude: -0.430

depth - mag: 0.387

depth - rms: 0.323

```
conn = sqlite3.connect('../data/earthquake_data.db')
cleaned_data.to_sql('earthquakes', conn, if_exists='replace',
index=False)
print("Data successfully loaded into SQLite database:
earthquake_data.db")
```

```
queries = {
    "Daily Summary": """
        SELECT date,
            COUNT(*) as earthquake_count,
            ROUND(AVG(mag), 2) as avg_magnitude,
            MAX(mag) as max_magnitude,
            ROUND(AVG(depth), 1) as avg_depth
        FROM earthquakes
        GROUP BY date
        ORDER BY date
    """,

    "Geographic Hotspots": """
        SELECT place,
            COUNT(*) as earthquake_count,
            ROUND(AVG(mag), 2) as avg_magnitude,
            MAX(mag) as max_magnitude
        FROM earthquakes
        GROUP BY place
        HAVING earthquake_count > 1
        ORDER BY earthquake_count DESC
        LIMIT 10
    """,

    "Network Performance": """
        SELECT net,
            COUNT(*) as earthquake_count,
            ROUND(AVG(mag), 2) as avg_magnitude,
            ROUND(AVG(magError), 3) as avg_mag_error
        FROM earthquakes
        GROUP BY net
        ORDER BY earthquake_count DESC
    ""
}

for query_name, query in queries.items():
    print(f"\n{query_name.upper()}")
    result = pd.read_sql_query(query, conn)
    print(result)

conn.close()
```


Data successfully loaded into SQLite database: earthquake_data.db

DAILY SUMMARY

	date	earthquake_count	avg_magnitude	max_magnitude
avg_depth				
0	1/1/2024	432	1.75	7.5
21.7				
1	1/10/2024	243	1.47	5.0
19.1				
2	1/11/2024	229	1.58	6.4
28.4				
3	1/12/2024	222	1.74	5.9
22.0				
4	1/13/2024	280	1.64	5.2
18.8				
5	1/14/2024	242	1.50	5.2
28.0				
6	1/15/2024	285	1.58	5.2
24.0				
7	1/16/2024	267	1.45	5.0
20.0				
8	1/17/2024	262	1.56	5.6
19.5				
9	1/18/2024	316	1.45	6.4
18.1				
10	1/19/2024	343	1.55	5.6
16.9				
11	1/2/2024	313	1.66	5.4
19.0				
12	1/20/2024	442	1.56	6.6
16.4				
13	1/21/2024	309	1.52	5.4
17.9				
14	1/22/2024	297	1.77	7.0
16.7				
15	1/23/2024	284	1.77	6.3
19.1				
16	1/24/2024	348	1.55	5.4
15.7				
17	1/25/2024	323	1.38	5.6
15.0				
18	1/26/2024	49	1.87	5.1
20.0				
19	1/3/2024	280	1.87	5.5
22.6				
20	1/4/2024	353	1.65	5.6
28.6				
21	1/5/2024	302	1.71	5.2
17.1				

22	1/6/2024	279	1.52	5.6
18.6				
23	1/7/2024	216	1.92	5.6
17.9				
24	1/8/2024	257	1.66	6.7
17.4				
25	1/9/2024	265	1.52	5.9
23.0				
26	12/27/2023	262	1.90	5.8
24.8				
27	12/28/2023	397	1.73	6.5
31.7				
28	12/29/2023	427	1.59	5.7
26.0				
29	12/30/2023	512	1.72	6.3
23.4				
30	12/31/2023	415	1.64	5.3
28.4				

GEOGRAPHIC HOTSPOTS

		place	earthquake_count
avg_magnitude	max_magnitude		
0	8 km NW of The Geysers, CA		140
0.84	2.83		
1	7 km NW of The Geysers, CA		106
0.94	2.23		
2	6 km NW of The Geysers, CA		81
0.94	3.41		
3	3 km NNE of Reno, Nevada		69
0.51	1.40		
4	7 km SW of Volcano, Hawaii		56
1.81	2.55		
5	7 km WNW of Cobb, CA		53
0.77	1.74		
6	29 km ESE of West Yellowstone, Montana		52
1.18	3.29		
7	10 km WNW of The Geysers, CA		51
0.77	1.53		
8	54 km NW of Toyah, Texas		49
1.95	3.00		
9	2 km NNE of Reno, Nevada		49
0.91	1.90		

NETWORK PERFORMANCE

	net	earthquake_count	avg_magnitude	avg_mag_error
0	ak	2132	1.66	0.173
1	nc	1602	0.98	0.203
2	ci	1063	1.18	0.148
3	us	1053	4.31	0.111
4	nn	642	0.67	0.287

5	tx	623	1.91	0.167
6	hv	581	1.94	0.867
7	av	563	0.14	0.250
8	uu	312	0.74	0.296
9	pr	223	2.72	0.107
10	ok	220	1.41	0.211
11	mb	214	1.23	0.203
12	uw	193	1.12	0.150
13	nm	24	1.94	0.097
14	se	6	2.07	0.075

```
print("EARTHQUAKE DATA ANALYSIS SUMMARY")
```

```
print(f"DATASET OVERVIEW:")
```

```
print(f"    • Time Period: {cleaned_data['date'].min()} to  
{cleaned_data['date'].max()}")
```

```
print(f"    • Total Earthquakes: {len(cleaned_data):,}")
```

```
print(f"    • Geographic Coverage: USA (Lat:  
{cleaned_data['latitude'].min():.1f}° to  
{cleaned_data['latitude'].max():.1f}°)")
```

```
print(f"    • Monitoring Networks: {cleaned_data['net'].nunique()}  
networks")
```

```
print(f"\nMAGNITUDE INSIGHTS:")
```

```
print(f"    • Magnitude Range: {cleaned_data['mag'].min():.1f} -  
{cleaned_data['mag'].max():.1f}")
```

```
print(f"    • Average Magnitude: {cleaned_data['mag'].mean():.2f}")
```

```
print(f"    • Most Common Mag Type: {cleaned_data['magType'].mode()[0]}  
(({cleaned_data['magType'] == cleaned_data['magType'].mode()  
[0]).sum()} earthquakes)")
```

```
print(f"    • High Magnitude (>4.0): {(cleaned_data['mag'] >  
4.0).sum()} earthquakes")
```

```
most_active_place = cleaned_data['place'].value_counts().index[0]
```

```
most_active_count = cleaned_data['place'].value_counts().iloc[0]
```

```
print(f"\nGEOGRAPHIC INSIGHTS:")
```

```
print(f"    • Most Active Location: {most_active_place}  
({most_active_count} earthquakes)")
```

```
print(f"    • Average Depth: {cleaned_data['depth'].mean():.1f} km")
```

```
print(f"    • Deepest Earthquake: {cleaned_data['depth'].max():.1f}  
km")
```

```
print(f"    • Shallowest Earthquake: {cleaned_data['depth'].min():.1f}  
km")
```

```
print(f"\nDATA QUALITY:")
```

```
print(f"    • Completeness: (((1 - cleaned_data.isnull().sum().sum() /  
(len(cleaned_data) * len(cleaned_data.columns))) * 100):.1f}%")
```

```
print(f"    • Average Magnitude Error:  
{cleaned_data['magError'].mean():.3f}")
```

```
print(f"    • Average Horizontal Error:
{cleaned_data['horizontalError'].mean():.2f} km")
```

EARTHQUAKE DATA ANALYSIS SUMMARY

DATASET OVERVIEW:

- Time Period: 1/1/2024 to 12/31/2023
- Total Earthquakes: 9,451
- Geographic Coverage: USA (Lat: -65.4° to 86.5°)
- Monitoring Networks: 15 networks

MAGNITUDE INSIGHTS:

- Magnitude Range: -1.3 - 7.5
- Average Magnitude: 1.63
- Most Common Mag Type: ml (6092 earthquakes)
- High Magnitude (>4.0): 880 earthquakes

GEOGRAPHIC INSIGHTS:

- Most Active Location: 8 km NW of The Geysers, CA (140 earthquakes)
- Average Depth: 21.4 km
- Deepest Earthquake: 660.8 km
- Shallowest Earthquake: -3.2 km

DATA QUALITY:

- Completeness: 100.0%
- Average Magnitude Error: 0.226
- Average Horizontal Error: 1.63 km