1. Overview of the Cleaned Dataset

	measure	location	sex	age	cause	metric	VdI
36	Deaths	Republic of Rwanda	Male	10 - 54 years	Other infectious diseases	Number	357.011761
37	Deaths	Republic of Rwanda	Female	10 - 54 years	Other infectious diseases	Number	243.007558
38	Deaths	Republic of Rwanda	Male	10 - 54 years	Other infectious diseases	Percent	0.023496
39	Deaths	Republic of Rwanda	Female	10 - 54 years	Other infectious diseases	Percent	0.021025
40	Deaths	Republic of Rwanda	Male	10 - 54 years	Other infectious diseases	Rate	8.323896
1073	Deaths	Republic of Burundi	Female	55+ years	Non-communicable diseases	Rate	2309.699600
1146	Deaths	Republic of Uganda	Male	<5 years	Other infectious diseases	Number	2704.302360
1147	Deaths	Republic of Uganda	Female	<5 years	Other infectious diseases	Number	1996.131428
1148	Deaths	Republic of Uganda	Male	<5 years	Other infectious diseases	Percent	0.049032
1149	Deaths	Republic of Uganda	Female	<5 years	Other infectious diseases	Percent	0.045008

250 rows x 7 columns

2. Top 5 Causes of Death Chart

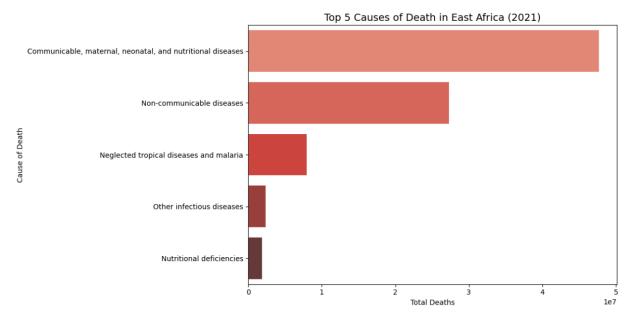
```
# Top 5 causes of death by number

top_causes = (
    df_2021[df_2021['metric'] == 'Number']
    .groupby('cause')['val']
    .sum()
    .sort_values(ascending=False)
    .head(5)
)

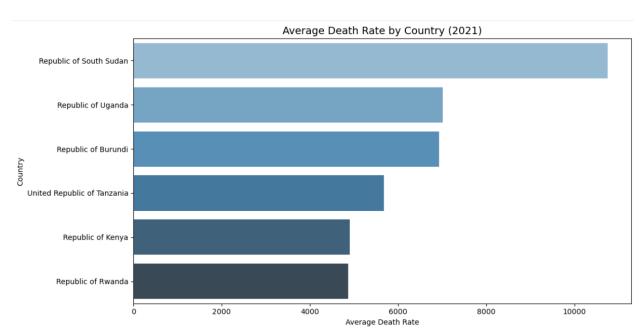
top_causes
```

cause

```
Communicable, maternal, neonatal, and nutritional diseases 4.772278e+07
Non-communicable diseases 2.730798e+07
Neglected tropical diseases and malaria 7.973004e+06
Other infectious diseases 2.344784e+06
Nutritional deficiencies 1.880506e+06
Name: val, dtype: float64
```



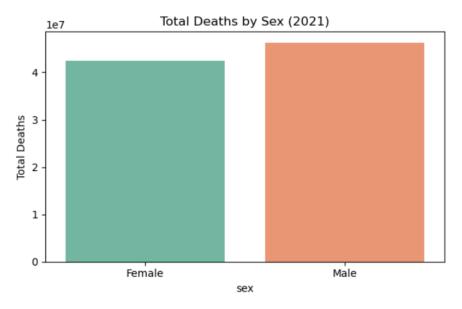
Death Rate by Country Chart



5. Deaths by Sex

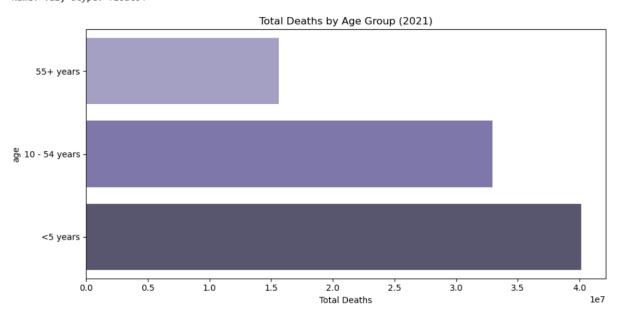
```
#Deaths by sex
sex_deaths = (
    df_2021[df_2021['metric'] == 'Number']
    .groupby('sex')['val']
    .sum()
)
sex_deaths

sex
Female    4.242193e+07
Male    4.626315e+07
Name: val, dtype: float64
```



6. Deaths by Age Group

age 55+ years 1.562700e+07 10 - 54 years 3.293644e+07 <5 years 4.012164e+07 Name: val, dtype: float64



7. Clustering Result

```
# MACHINE LEARNING - CLUSTERING

# Pivot table for clustering: Country vs Cause death rates

df_cluster = df_2021[df_2021['metric'] == 'Rate']
pivot = df_cluster.pivot_table(index='location', columns='cause', values='val', fill_value=0)

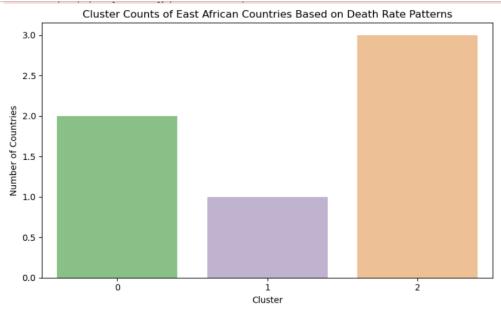
# Scale features
scaler = StandardScaler()
X = scaler.fit_transform(pivot)

# Fit KMeans Clustering
kmeans = KMeans(n_clusters=3, random_state=42)
clusters = kmeans.fit_predict(X)

# Assign clusters back to country
pivot['Cluster'] = clusters

# Evaluate clustering
score = silhouette_score(X, clusters)
print(f"\n \subseteq Silhouette Score: {round(score, 2)} (higher is better, max=1)\n")
```

☑ Silhouette Score: 0.18 (higher is better, max=1)



Ocuntry Cluster Assignments:

location
Republic of Kenya 0
Republic of Rwanda 0
Republic of South Sudan 1
Republic of Burundi 2
Republic of Uganda 2
United Republic of Tanzania 2
Name: Cluster, dtype: int32

8. Cause Diversity vs Avg Rate (Innovation Visual)

```
# Calculate cause diversity (number of non-zero causes per country)
pivot['Cause\ Diversity'] = (pivot.drop('Cluster',\ axis=1) \ > \ 0).sum(axis=1)
# Add average death rate per country
pivot['Average Rate'] = pivot.drop(columns=['Cluster', 'Cause Diversity']).mean(axis=1)
# Plot using seaborn with explicit data frame
plt.figure(figsize=(10, 5))
sns.scatterplot(
   data=pivot,
   x='Cause Diversity',
   y='Average Rate',
    hue='Cluster',
    palette='Set1'
plt.title('Cause Diversity vs. Average Death Rate (per Country)')
plt.xlabel('Number of Causes with Non-Zero Rate')
plt.ylabel('Average Death Rate')
plt.tight_layout()
plt.show()
```

