

# globalmisssingmigrants-1

January 5, 2024

## *Global Missing Migrants*

This dataset presents a solemn record of individuals who embarked on perilous journeys towards international destinations, only to go missing or tragically lose their lives along the way. The dataset is a result of the ongoing efforts of the Missing Migrants Project, an initiative by the International Organization for Migration (IOM) since 2014.

Migration is a complex and multifaceted phenomenon that touches the lives of millions of people worldwide. This dataset sheds light on the challenges faced by migrants, as well as the immense courage and resilience they display. While the numbers presented here offer a glimpse into the scope of the issue, it's important to acknowledge that the true extent of the problem is likely underestimated due to the inherent difficulties in collecting such data.

The data here shows details such as the date the migrants went missing, the number of migrants that went missing, the region in which the incident occurred, etc. It has the potential to be very helpful in determining the severity of the issue of missing migrants in different regions across the world.

### FEATURES:

Incident Type: Type of migration incident

Incident Year: Year when the incident occurred

Reported Month: Month when the incident was reported

Region of Origin: Geographical region where the migrants originated

Region of Incident: Geographical region where the incident occurred

Country of Origin: Country from which the migrants originated

Number of Dead: Number of confirmed deceased migrants

Minimum Estimated Number of Missing: Minimum estimated count of missing migrants

Total Number of Dead and Missing: Total count of both deceased and missing migrants

Number of Survivors: Number of migrants who survived the incident

Number of Females: Number of female migrants involved

Number of Males: Number of male migrants involved

Number of Children: Number of children migrants involved

Cause of Death: Cause of death for the migrants

## GOAL

To predict the Total Number of Dead and Missing

```
[ ]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib import figure
df=pd.read_csv('/content/Global Missing Migrants Dataset.csv.zip')
df
```

```
[ ]: Incident Type Incident year Reported Month \
0 Incident 2014 January
1 Incident 2014 January
2 Incident 2014 January
3 Incident 2014 January
4 Incident 2014 January
...
13015 Incident 2023 July
13016 Incident 2023 July
13017 Incident 2023 July
13018 Incident 2023 July
13019 Incident 2023 July
```

```
Region of Origin Region of Incident Country of Origin \
0 Central America North America Guatemala
1 Latin America / Caribbean (P) North America Unknown
2 Latin America / Caribbean (P) North America Unknown
3 Central America North America Mexico
4 Northern Africa Europe Sudan
...
13015 Western Asia Western Asia Syrian Arab Republic
13016 Western Africa (P) Western Asia Unknown
13017 Western Africa Northern Africa Senegal
13018 Mixed Northern Africa Unknown
13019 Western Africa (P) Western Africa Unknown
```

```
Number of Dead Minimum Estimated Number of Missing \
0 1.0 0
1 1.0 0
2 1.0 0
3 1.0 0
4 1.0 0
...
13015 4.0 0
13016 2.0 0
13017 13.0 0
```

13018	6.0	0
13019	16.0	37

	Total Number of Dead and Missing	Number of Survivors \
0	1	0
1	1	0
2	1	0
3	1	0
4	1	2
...	...	...
13015	4	0
13016	2	0
13017	13	6
13018	6	48
13019	53	2

	Number of Females	Number of Males	Number of Children \
0	0	1	0
1	0	0	0
2	0	0	0
3	0	1	0
4	0	1	0
...	...	...	...
13015	0	4	0
13016	0	2	0
13017	0	0	0
13018	0	0	0
13019	2	0	0

	Cause of Death \
0	Mixed or unknown
1	Mixed or unknown
2	Mixed or unknown
3	Violence
4	Harsh environmental conditions / lack of adequ...
...	...
13015	Vehicle accident / death linked to hazardous t...
13016	Vehicle accident / death linked to hazardous t...
13017	Drowning
13018	Drowning
13019	Drowning

	Migration route \
0	US-Mexico border crossing
1	US-Mexico border crossing
2	US-Mexico border crossing
3	US-Mexico border crossing

4		NaN
...		...
13015	Türkiye-Europe land route	
13016	Türkiye-Europe land route	
13017	Western Africa / Atlantic route to the Canary ...	
13018	Western Africa / Atlantic route to the Canary ...	
13019	Western Africa / Atlantic route to the Canary ...	

	Location of death \
0	Pima Country Office of the Medical Examiner ju...
1	Pima Country Office of the Medical Examiner ju...
2	Pima Country Office of the Medical Examiner ju...
3	near Douglas, Arizona, USA
4	Border between Russia and Estonia
...	...
13015	In Ipsala, Edirne province, Türkiye - travelli...
13016	At the Kapıkule Türkiye-Bulgaria Border Gate, ...
13017	Off the coasts of Dakhla, Western Sahara - 6 s...
13018	Unspecified location off the coast of Nador, M...
13019	Off the coast of Ouakam, Dakar, Senegal

	Information Source \
0	Pima County Office of the Medical Examiner (PC...
1	Pima County Office of the Medical Examiner (PC...
2	Pima County Office of the Medical Examiner (PC...
3	Ministry of Foreign Affairs Mexico, Pima Count...
4	EUBusiness (Agence France-Presse)
...	...
13015	Andalou Agency, Son Dakika, Orient News
13016	Son Dakika, Hurriyet
13017	Barron's News, InfoMigrants, IOM Morrocco
13018	El Nashra, Swiss Info; CGTN, IOM Morrocco
13019	IOM Senegal

	Coordinates	UNSD Geographical Grouping
0	31.650259, -110.366453	Northern America
1	31.59713, -111.73756	Northern America
2	31.94026, -113.01125	Northern America
3	31.506777, -109.315632	Northern America
4	59.1551, 28	Northern Europe
...	...	...
13015	40.91271268, 26.369657	Western Asia
13016	41.71697242, 26.351489	Western Asia
13017	23.72836078, -15.901632	Uncategorized
13018	35.17187365, -2.903182	Uncategorized
13019	14.71870705, -17.506255	Uncategorized

[13020 rows x 19 columns]

```
[ ]: df.head()
```

```
[ ]: Incident Type Incident year Reported Month Region of Origin \
0 Incident 2014 January Central America
1 Incident 2014 January Latin America / Caribbean (P)
2 Incident 2014 January Latin America / Caribbean (P)
3 Incident 2014 January Central America
4 Incident 2014 January Northern Africa
```

```
Region of Incident Country of Origin Number of Dead \
0 North America Guatemala 1.0
1 North America Unknown 1.0
2 North America Unknown 1.0
3 North America Mexico 1.0
4 Europe Sudan 1.0
```

```
Minimum Estimated Number of Missing Total Number of Dead and Missing \
0 0 1
1 0 1
2 0 1
3 0 1
4 0 1
```

```
Number of Survivors Number of Females Number of Males \
0 0 0 1
1 0 0 0
2 0 0 0
3 0 0 1
4 2 0 1
```

```
Number of Children Cause of Death \
0 0 Mixed or unknown
1 0 Mixed or unknown
2 0 Mixed or unknown
3 0 Violence
4 0 Harsh environmental conditions / lack of adequ...
```

```
Migration route \
0 US-Mexico border crossing
1 US-Mexico border crossing
2 US-Mexico border crossing
3 US-Mexico border crossing
4 NaN
```

```
Location of death \
```

```

0 Pima Country Office of the Medical Examiner ju...
1 Pima Country Office of the Medical Examiner ju...
2 Pima Country Office of the Medical Examiner ju...
3      near Douglas, Arizona, USA
4      Border between Russia and Estonia

```

```

                                Information Source      Coordinates \
0 Pima County Office of the Medical Examiner (PC... 31.650259, -110.366453
1 Pima County Office of the Medical Examiner (PC... 31.59713, -111.73756
2 Pima County Office of the Medical Examiner (PC... 31.94026, -113.01125
3 Ministry of Foreign Affairs Mexico, Pima Count... 31.506777, -109.315632
4      EUBusiness (Agence France-Presse)      59.1551, 28

```

```

UNSD Geographical Grouping
0      Northern America
1      Northern America
2      Northern America
3      Northern America
4      Northern Europe

```

```
[ ]: df.tail()
```

```

[ ]:      Incident Type  Incident year Reported Month      Region of Origin \
13015      Incident      2023      July      Western Asia
13016      Incident      2023      July  Western Africa (P)
13017      Incident      2023      July      Western Africa
13018      Incident      2023      July      Mixed
13019      Incident      2023      July  Western Africa (P)

```

```

      Region of Incident      Country of Origin  Number of Dead \
13015      Western Asia  Syrian Arab Republic      4.0
13016      Western Asia      Unknown      2.0
13017      Northern Africa      Senegal      13.0
13018      Northern Africa      Unknown      6.0
13019      Western Africa      Unknown      16.0

```

```

      Minimum Estimated Number of Missing  Total Number of Dead and Missing \
13015      0      4
13016      0      2
13017      0      13
13018      0      6
13019      37      53

```

```

      Number of Survivors  Number of Females  Number of Males \
13015      0      0      4
13016      0      0      2
13017      6      0      0

```

13018	48	0	0
13019	2	2	0

	Number of Children	Cause of Death \
13015	0	Vehicle accident / death linked to hazardous t...
13016	0	Vehicle accident / death linked to hazardous t...
13017	0	Drowning
13018	0	Drowning
13019	0	Drowning

	Migration route \
13015	Türkiye-Europe land route
13016	Türkiye-Europe land route
13017	Western Africa / Atlantic route to the Canary ...
13018	Western Africa / Atlantic route to the Canary ...
13019	Western Africa / Atlantic route to the Canary ...

	Location of death \
13015	In Ipsala, Edirne province, Türkiye - travelli...
13016	At the Kapıkule Türkiye-Bulgaria Border Gate, ...
13017	Off the coasts of Dakhla, Western Sahara - 6 s...
13018	Unspecified location off the coast of Nador, M...
13019	Off the coast of Ouakam, Dakar, Senegal

	Information Source	Coordinates \
13015	Andalou Agency, Son Dakika, Orient News	40.91271268, 26.369657
13016	Son Dakika, Hurriyet	41.71697242, 26.351489
13017	Barron's News, InfoMigrants, IOM Morrocco	23.72836078, -15.901632
13018	El Nashra, Swiss Info; CGTN, IOM Morrocco	35.17187365, -2.903182
13019	IOM Senegal	14.71870705, -17.506255

	UNSD Geographical Grouping
13015	Western Asia
13016	Western Asia
13017	Uncategorized
13018	Uncategorized
13019	Uncategorized

```
[ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13020 entries, 0 to 13019
Data columns (total 19 columns):
```

#	Column	Non-Null Count	Dtype
0	Incident Type	13020 non-null	object
1	Incident year	13020 non-null	int64

2	Reported Month	13020	non-null	object
3	Region of Origin	12998	non-null	object
4	Region of Incident	13020	non-null	object
5	Country of Origin	13012	non-null	object
6	Number of Dead	12470	non-null	float64
7	Minimum Estimated Number of Missing	13020	non-null	int64
8	Total Number of Dead and Missing	13020	non-null	int64
9	Number of Survivors	13020	non-null	int64
10	Number of Females	13020	non-null	int64
11	Number of Males	13020	non-null	int64
12	Number of Children	13020	non-null	int64
13	Cause of Death	13020	non-null	object
14	Migration route	9999	non-null	object
15	Location of death	13020	non-null	object
16	Information Source	13012	non-null	object
17	Coordinates	12984	non-null	object
18	UNSD Geographical Grouping	13019	non-null	object

dtypes: float64(1), int64(7), object(11)  
memory usage: 1.9+ MB

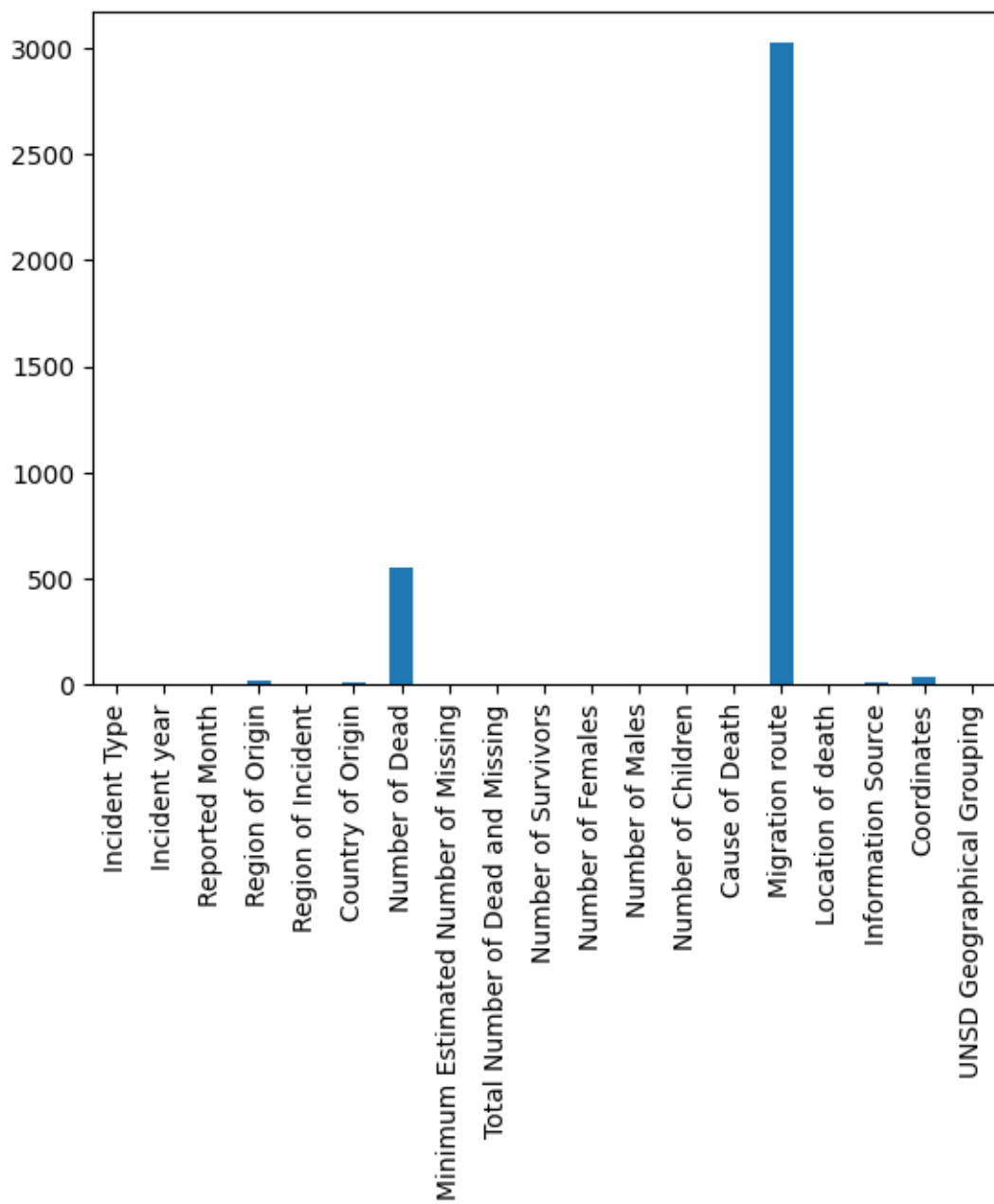
```
[ ]: df.isna().sum()
```

```
[ ]: Incident Type          0
Incident year              0
Reported Month            0
Region of Origin          22
Region of Incident        0
Country of Origin         8
Number of Dead            550
Minimum Estimated Number of Missing  0
Total Number of Dead and Missing  0
Number of Survivors       0
Number of Females         0
Number of Males           0
Number of Children        0
Cause of Death            0
Migration route           3021
Location of death         0
Information Source         8
Coordinates               36
UNSD Geographical Grouping  1
dtype: int64
```

```
[ ]: df.isna().sum().plot(kind='bar')
```

```
[ ]: <Axes: >
```





```
[ ]: df.dtypes
```

```
[ ]: Incident Type          object
      Incident year        int64
      Reported Month       object
      Region of Origin     object
      Region of Incident   object
      Country of Origin    object
```

Number of Dead	float64
Minimum Estimated Number of Missing	int64
Total Number of Dead and Missing	int64
Number of Survivors	int64
Number of Females	int64
Number of Males	int64
Number of Children	int64
Cause of Death	object
Migration route	object
Location of death	object
Information Source	object
Coordinates	object
UNSD Geographical Grouping	object
dtype:	object

### *Visualization of object Columns*

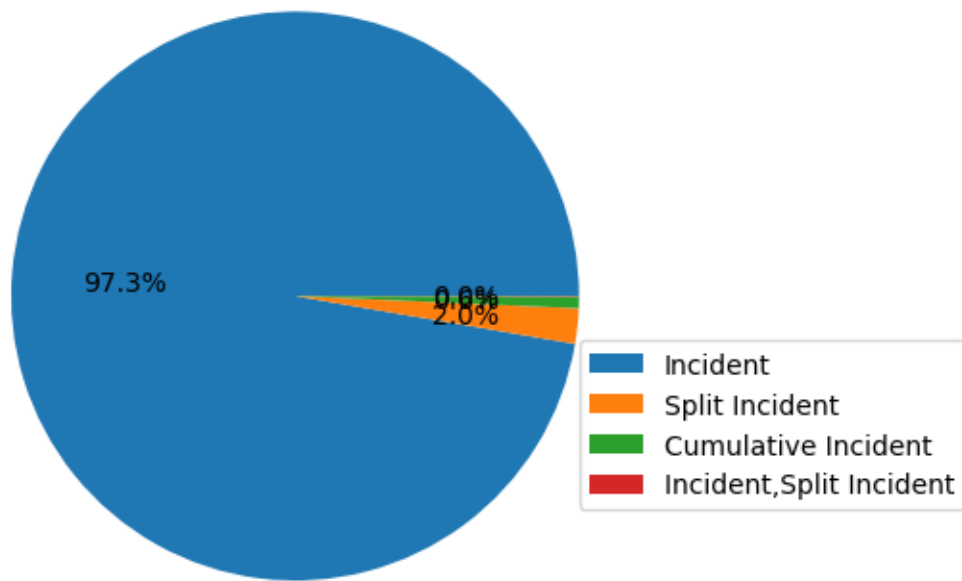
```
[ ]: incident=df['Incident Type'].value_counts()
incident
```

```
[ ]: Incident          12670
Split Incident         261
Cumulative Incident     84
Incident,Split Incident    5
Name: Incident Type, dtype: int64
```

```
[ ]: plt.pie(incident,autopct='%2.1f%%')
plt.legend(incident.index,loc=(0.9,0.2))
plt.title("Incident Type")
```

```
[ ]: Text(0.5, 1.0, 'Incident Type')
```

### Incident Type

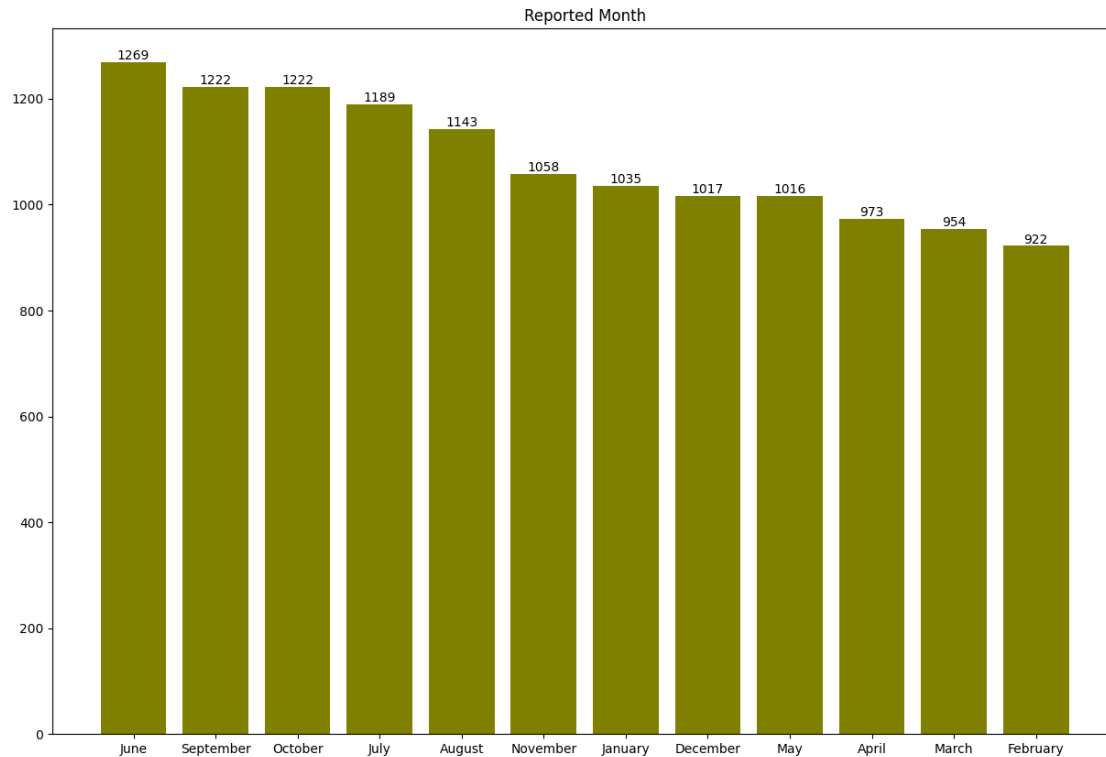


```
[ ]: Report=df['Reported Month'].value_counts()
Report
```

```
[ ]: June          1269
      September    1222
      October      1222
      July          1189
      August        1143
      November     1058
      January       1035
      December     1017
      May           1016
      April          973
      March          954
      February      922
      Name: Reported Month, dtype: int64
```

```
[ ]: plt.figure(figsize=(15,10))
      plt.bar(Report.index,Report,color="olive")
      for i,count in enumerate(Report):
          plt.text(Report.index[i],count,str(count),ha='center',va='bottom')
      plt.title('Reported Month')
```

```
[ ]: Text(0.5, 1.0, 'Reported Month')
```



```
[ ]: Origin=df['Region of Origin'].value_counts()
Origin
```

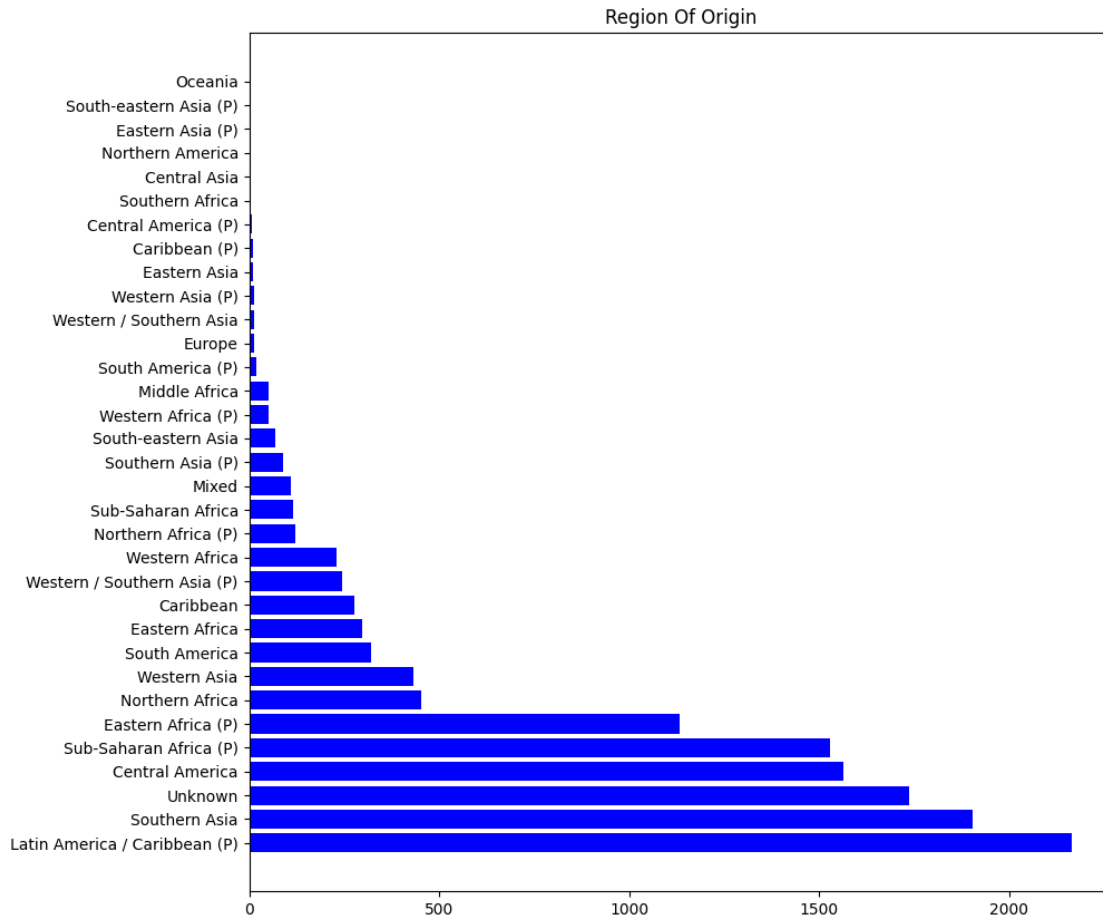
```
[ ]: Latin America / Caribbean (P)    2164
      Southern Asia                    1904
      Unknown                          1737
      Central America                  1565
      Sub-Saharan Africa (P)           1528
      Eastern Africa (P)               1133
      Northern Africa                  452
      Western Asia                     432
      South America                    322
      Eastern Africa                   298
      Caribbean                        278
      Western / Southern Asia (P)      245
      Western Africa                   229
      Northern Africa (P)              122
      Sub-Saharan Africa               116
      Mixed                            111
      Southern Asia (P)                 90
      South-eastern Asia                68
```

Western Africa (P)	52
Middle Africa	51
South America (P)	18
Europe	14
Western / Southern Asia	14
Western Asia (P)	12
Eastern Asia	11
Caribbean (P)	10
Central America (P)	8
Southern Africa	5
Central Asia	4
Northern America	2
Eastern Asia (P)	1
South-eastern Asia (P)	1
Oceania	1

Name: Region of Origin, dtype: int64

```
[ ]: plt.figure(figsize=(10,10))
plt.barh(Origin.index,Origin,color="blue")
plt.title('Region Of Origin')
```

```
[ ]: Text(0.5, 1.0, 'Region Of Origin')
```



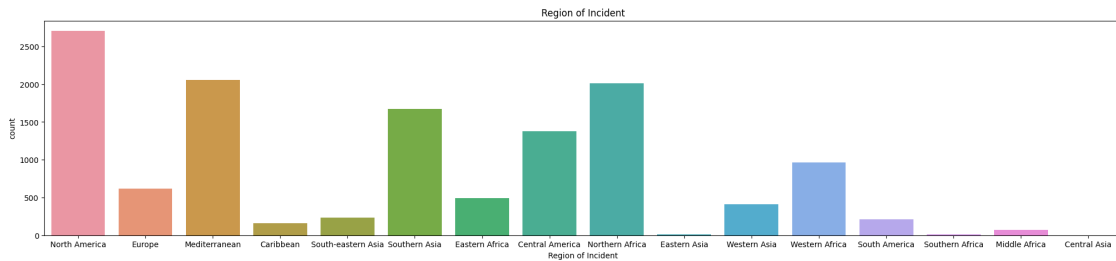
```
[ ]: Incident1=df["Region of Incident"].value_counts()
Incident1
```

```
[ ]: North America      2706
Mediterranean          2055
Northern Africa        2014
Southern Asia           1673
Central America         1375
Western Africa          967
Europe                  619
Eastern Africa          489
Western Asia            414
South-eastern Asia      237
South America           209
Caribbean              160
Middle Africa           75
Southern Africa         16
Eastern Asia            10
```

```
Central Asia          1
Name: Region of Incident, dtype: int64
```

```
[ ]: plt.figure(figsize=(25,5))
sns.countplot(x='Region of Incident',data=df)
plt.title("Region of Incident")
```

```
[ ]: Text(0.5, 1.0, 'Region of Incident')
```



```
[ ]: country=df['Country of Origin'].value_counts()
country
```

```
[ ]: Unknown          7220
Afghanistan          1702
Mexico               709
Syrian Arab Republic 308
Honduras             307

Nigeria,Sudan          1
Mali,Senegal,Unknown    1
Somalia,Unknown         1
Cameroon,Côte d'Ivoire,Democratic Republic of the Congo,Tunisia 1
Gambia,Mali,Nigeria,Senegal,Unknown 1
Name: Country of Origin, Length: 335, dtype: int64
```

```
[ ]: Death=df['Cause of Death'].value_counts()
Death
```

```
[ ]: Drowning          3313
Mixed or unknown       3175
Vehicle accident / death linked to hazardous transport 2112
Harsh environmental conditions / lack of adequate shelter, food, water 1360
Violence
```

```

1313
Sickness / lack of access to adequate healthcare
1219
Accidental death
507
Drowning,Harsh environmental conditions / lack of adequate shelter, food, water
8
Drowning,Mixed or unknown
4
Harsh environmental conditions / lack of adequate shelter, food, water,Sickness
/ lack of access to adequate healthcare      3
Drowning,Violence
2
Drowning,Vehicle accident / death linked to hazardous transport
1
Harsh environmental conditions / lack of adequate shelter, food, water,Mixed or
unknown      1
Mixed or unknown,Vehicle accident / death linked to hazardous transport,Violence
1
Drowning,Sickness / lack of access to adequate healthcare
1
Name: Cause of Death, dtype: int64

```

```

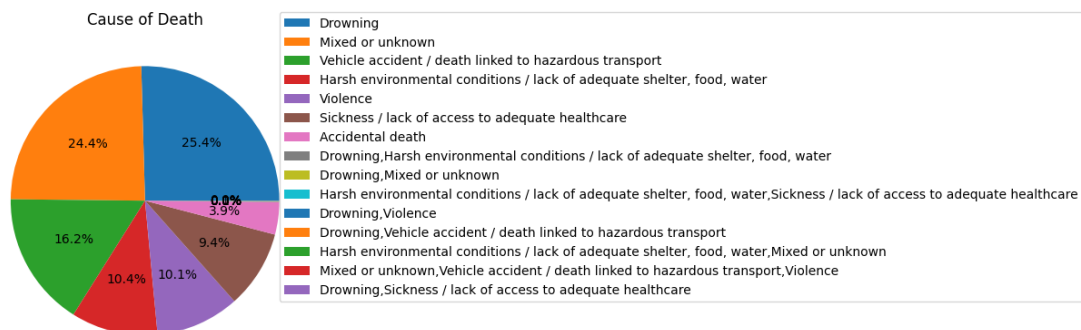
[ ]: plt.pie(Death,autopct='%1.1f%%')
plt.legend(Death.index,loc=(0.9,0.2))
plt.title("Cause of Death")

```

```

[ ]: Text(0.5, 1.0, 'Cause of Death')

```



```

[ ]: Migration=df['Migration route'].value_counts()
Migration

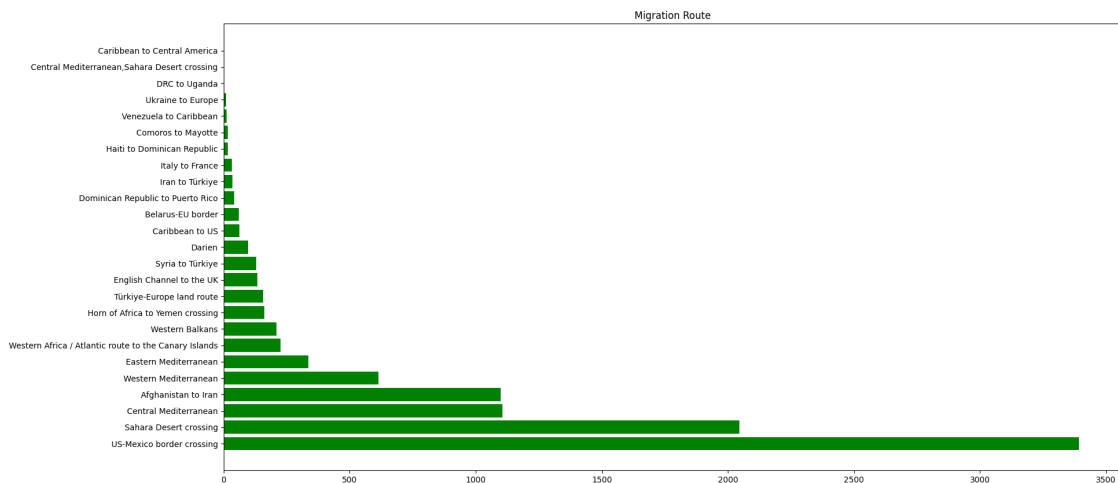
```



```
[ ]: US-Mexico border crossing 3392
Sahara Desert crossing 2046
Central Mediterranean 1106
Afghanistan to Iran 1099
Western Mediterranean 614
Eastern Mediterranean 336
Western Africa / Atlantic route to the Canary Islands 226
Western Balkans 210
Horn of Africa to Yemen crossing 161
Türkiye-Europe land route 157
English Channel to the UK 134
Syria to Türkiye 129
Darien 98
Caribbean to US 63
Belarus-EU border 61
Dominican Republic to Puerto Rico 42
Iran to Türkiye 34
Italy to France 33
Haiti to Dominican Republic 17
Comoros to Mayotte 16
Venezuela to Caribbean 11
Ukraine to Europe 9
DRC to Uganda 3
Central Mediterranean,Sahara Desert crossing 1
Caribbean to Central America 1
Name: Migration route, dtype: int64
```

```
[ ]: plt.figure(figsize=(20,10))
plt.barh(Migration.index,Migration,color="green")
plt.title("Migration Route")
```

```
[ ]: Text(0.5, 1.0, 'Migration Route')
```



```
[ ]: Location=df['Location of death'].value_counts()
Location
```

```
[ ]: Pima Country Office of the Medical Examiner jurisdiction, Arizona, USA (see
coordinates for exact location)      1061
Pima County Office of the Medical Examiner jurisdiction, Arizona, USA (see
coordinates for exact location)      404
Reported at Milak border crossing, Iran
200
Agadez, Niger
121
Sahara desert, Libya
116
...
Evros River, near Orestiad, Greece
1
Bodies recovered near Plage de Trougout, Nador, Morocco
1
Namanga, Tanzania-Kenya border
1
Ndola, Zambia, near border with Democratic Republic of the Congo
1
Off the coast of Ouakam, Dakar, Senegal
1
Name: Location of death, Length: 7460, dtype: int64
```

```
[ ]: Info=df["Information Source"].value_counts()
Info
```

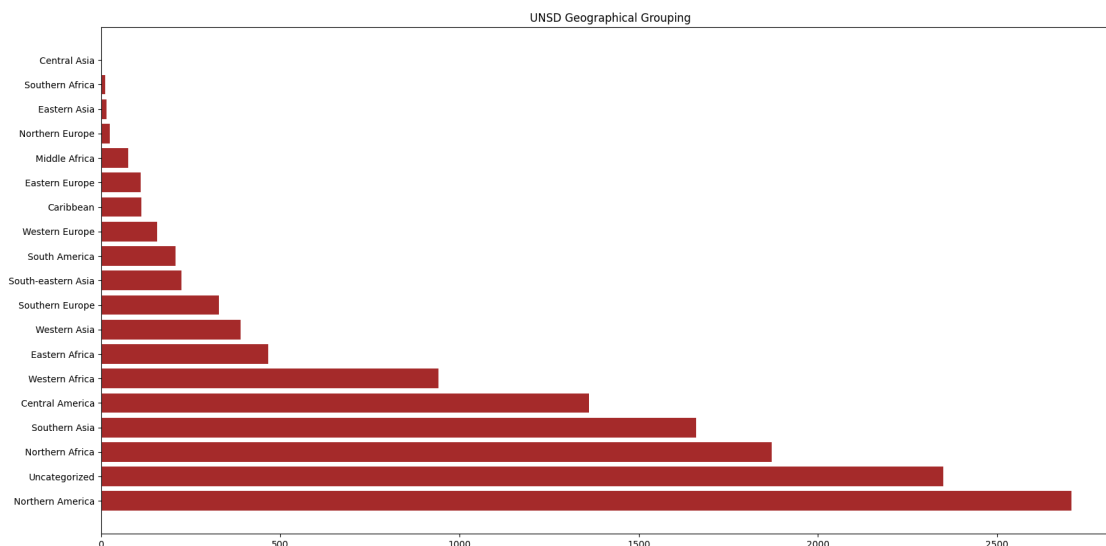
```
[ ]: IOM Afghanistan                                1538
Pima County Office of the Medical Examiner (PCOME)  1480
Mixed Migration Monitoring Mechanism Initiative (4Mi) 1089
Mixed Migration Monitoring Mechanism Initiative (4mi)  992
Mixed Migration Monitoring Mechanism Initiative (4mi)  673
...
El Faro de Ceuta, Europa Press                        1
Buzzfeed News, Al Jazeera                            1
Eagle Pass Texas News, Zócalo                        1
Posto                                                  1
El Nashra, Swiss Info; CGTN, IOM Morrocco            1
Name: Information Source, Length: 3803, dtype: int64
```

```
[ ]: Grouping=df['UNSD Geographical Grouping'].value_counts()
Grouping
```

```
[ ]: Northern America      2708
      Uncategorized      2351
      Northern Africa      1872
      Southern Asia      1660
      Central America      1362
      Western Africa      941
      Eastern Africa      467
      Western Asia      389
      Southern Europe      329
      South-eastern Asia    225
      South America      207
      Western Europe      156
      Caribbean          113
      Eastern Europe      111
      Middle Africa       75
      Northern Europe      25
      Eastern Asia        15
      Southern Africa      12
      Central Asia         1
      Name: UNSD Geographical Grouping, dtype: int64
```

```
[ ]: plt.figure(figsize=(20,10))
      plt.barh(Grouping.index,Grouping,color='brown')
      plt.title("UNSD Geographical Grouping")
```

```
[ ]: Text(0.5, 1.0, 'UNSD Geographical Grouping')
```



*Filling*

```
[ ]: df['Number of Dead'].unique()
```

```
[ ]: array([ 1., 12., 5., 15., 2., 8., 11., 7., 251., 17., 10.,
          4., 0., 6., 22., 44., 13., 62., 3., 9., 45., 29.,
          20., 170., 18., 24., 42., 64., 70., 41., 27., 21., 16.,
          111., nan, 26., 750., 14., 36., 47., 106., 30., 100., 40.,
          49., 52., 71., 37., 61., 34., 95., 28., 43., 57., 19.,
          23., 123., 35., 39., 25., 51., 133., 120., 204., 97., 87.,
          54., 32., 74., 33., 31., 48., 84., 46., 38., 83., 75.,
          53., 55., 167., 56., 50., 117., 160., 60., 86., 80.]
```

```
[ ]: ##Filling missing Values
df['Number of Dead'].fillna(0, inplace=True)
```

```
[ ]: df['Region of Origin'].fillna('Unknown', inplace=True)
df['Country of Origin'].fillna('Unknown', inplace=True)
df['Migration route'].fillna('Unknown', inplace=True)
df['Information Source'].fillna('Unknown', inplace=True)
df['UNSD Geographical Grouping'].fillna('Unknown', inplace=True)
df['Coordinates']=df['Coordinates'].fillna(df['Coordinates'].mode()[0])
```

```
[ ]: df.isna().sum()
```

```
[ ]: Incident Type                0
Incident year                    0
Reported Month                   0
Region of Origin                 0
Region of Incident               0
Country of Origin                0
Number of Dead                   0
Minimum Estimated Number of Missing 0
Total Number of Dead and Missing 0
Number of Survivors              0
Number of Females                0
Number of Males                  0
Number of Children               0
Cause of Death                   0
Migration route                  0
Location of death                0
Information Source               0
Coordinates                      0
UNSD Geographical Grouping       0
dtype: int64
```

### *Encoding*

```
[ ]: from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
df['Incident Type']=le.fit_transform(df['Incident Type'])
df['Reported Month']=le.fit_transform(df['Reported Month'])
df['Region of Origin']=le.fit_transform(df['Region of Origin'])
df['Region of Incident']=le.fit_transform(df['Region of Incident'])
df['Country of Origin']=le.fit_transform(df['Country of Origin'])
df['Cause of Death']=le.fit_transform(df['Cause of Death'])
df['Migration route']=le.fit_transform(df['Migration route'])
df['Location of death']=le.fit_transform(df['Location of death'])
df['Information Source']=le.fit_transform(df['Information Source'])
df['UNSD Geographical Grouping']=le.fit_transform(df['UNSD Geographical_
↳Grouping'])

[ ]: df[['Latitude', 'Longitude']] = df['Coordinates'].str.split(',', expand=True).
↳astype(float)
df
```

```
[ ]:
```

	Incident Type	Incident year	Reported Month	Region of Origin \
0	1	2014	4	2
1	1	2014	4	10
2	1	2014	4	10
3	1	2014	4	2
4	1	2014	4	13
...	...	...	...	...
13015	1	2023	5	31
13016	1	2023	5	30
13017	1	2023	5	29
13018	1	2023	5	12
13019	1	2023	5	30

	Region of Incident	Country of Origin	Number of Dead \
0	8	195	1.0
1	8	326	1.0
2	8	326	1.0
3	8	259	1.0
4	5	315	1.0
...	...	...	...
13015	15	318	4.0
13016	15	326	2.0
13017	9	303	13.0
13018	9	326	6.0
13019	14	326	16.0

	Minimum Estimated Number of Missing	Total Number of Dead and Missing \
0	0	1
1	0	1

2	0	1
3	0	1
4	0	1
...	...	...
13015	0	4
13016	0	2
13017	0	13
13018	0	6
13019	37	53

	Number of Survivors	...	Number of Males	Number of Children	\
0	0	...	1	0	
1	0	...	0	0	
2	0	...	0	0	
3	0	...	1	0	
4	2	...	1	0	
...	...	...	...	...	
13015	0	...	4	0	
13016	0	...	2	0	
13017	6	...	0	0	
13018	48	...	0	0	
13019	2	...	0	0	

	Cause of Death	Migration route	Location of death	Information Source	\
0	10	19	4562	2784	
1	10	19	4562	2784	
2	10	19	4562	2784	
3	14	19	7373	2521	
4	7	21	1362	890	
...	...	...	...	...	
13015	13	18	2506	277	
13016	13	18	678	3127	
13017	1	23	4281	415	
13018	1	23	7056	1114	
13019	1	23	4165	1840	

	Coordinates	UNSD Geographical Grouping	Latitude	\
0	31.650259, -110.366453	8	31.650259	
1	31.59713, -111.73756	8	31.597130	
2	31.94026, -113.01125	8	31.940260	
3	31.506777, -109.315632	8	31.506777	
4	59.1551, 28	9	59.155100	
...	...	...	...	
13015	40.91271268, 26.369657	18	40.912713	
13016	41.71697242, 26.351489	18	41.716972	
13017	23.72836078, -15.901632	15	23.728361	
13018	35.17187365, -2.903182	15	35.171874	

13019 14.71870705, -17.506255

15 14.718707

```

    Longitude
0    -110.366453
1    -111.737560
2    -113.011250
3    -109.315632
4      28.000000
...
13015  26.369657
13016  26.351489
13017 -15.901632
13018  -2.903182
13019 -17.506255
```

[13020 rows x 21 columns]

```
[ ]: df.dtypes
```

```
[ ]: Incident Type                int64
     Incident year                int64
     Reported Month              int64
     Region of Origin            int64
     Region of Incident          int64
     Country of Origin           int64
     Number of Dead              float64
     Minimum Estimated Number of Missing int64
     Total Number of Dead and Missing int64
     Number of Survivors         int64
     Number of Females           int64
     Number of Males             int64
     Number of Children          int64
     Cause of Death              int64
     Migration route             int64
     Location of death           int64
     Information Source           int64
     Coordinates                 object
     UNSD Geographical Grouping  int64
     Latitude                   float64
     Longitude                   float64
     dtype: object
```

### *Correlation*

```
[ ]: df.corr()
```

<ipython-input-32-2f6f6606aa2c>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will

default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
df.corr()
```

```
[ ]:
```

	Incident Type	Incident year \
Incident Type	1.000000	0.076628
Incident year	0.076628	1.000000
Reported Month	-0.008714	-0.036519
Region of Origin	0.038305	0.055448
Region of Incident	-0.060845	0.063038
Country of Origin	-0.089466	-0.301706
Number of Dead	0.058494	-0.070679
Minimum Estimated Number of Missing	0.070826	-0.055558
Total Number of Dead and Missing	0.083497	-0.077348
Number of Survivors	0.109269	-0.044353
Number of Females	0.062180	-0.015575
Number of Males	0.148464	0.027849
Number of Children	0.027818	-0.030011
Cause of Death	-0.099913	-0.095503
Migration route	-0.055576	-0.099958
Location of death	-0.042855	0.029271
Information Source	-0.030788	-0.159215
UNSD Geographical Grouping	0.050894	0.052416
Latitude	0.043708	0.090723
Longitude	0.013392	0.006074

	Reported Month	Region of Origin \
Incident Type	-0.008714	0.038305
Incident year	-0.036519	0.055448
Reported Month	1.000000	0.035904
Region of Origin	0.035904	1.000000
Region of Incident	0.026476	0.447338
Country of Origin	-0.025246	-0.025367
Number of Dead	-0.019174	0.029656
Minimum Estimated Number of Missing	-0.007721	0.043659
Total Number of Dead and Missing	-0.015219	0.048411
Number of Survivors	-0.006513	0.076922
Number of Females	0.015927	0.056673
Number of Males	-0.024489	0.030489
Number of Children	-0.003678	0.045628
Cause of Death	0.020247	-0.028682
Migration route	0.006460	-0.261433
Location of death	-0.012376	0.072978
Information Source	-0.016946	-0.187777
UNSD Geographical Grouping	0.021245	0.617495
Latitude	-0.005279	0.172293
Longitude	0.044589	0.599155



	Region of Incident	Country of Origin \
Incident Type	-0.060845	-0.089466
Incident year	0.063038	-0.301706
Reported Month	0.026476	-0.025246
Region of Origin	0.447338	-0.025367
Region of Incident	1.000000	-0.226761
Country of Origin	-0.226761	1.000000
Number of Dead	-0.004398	0.001795
Minimum Estimated Number of Missing	-0.043498	0.002398
Total Number of Dead and Missing	-0.036213	0.002738
Number of Survivors	-0.042863	0.024192
Number of Females	-0.014861	0.026061
Number of Males	-0.001823	-0.032380
Number of Children	0.004213	0.001961
Cause of Death	0.283421	-0.011881
Migration route	-0.168905	0.385098
Location of death	0.045629	-0.110229
Information Source	0.023146	0.264236
UNSD Geographical Grouping	0.628374	-0.130228
Latitude	-0.060237	-0.187679
Longitude	0.469840	-0.294024

	Number of Dead \
Incident Type	0.058494
Incident year	-0.070679
Reported Month	-0.019174
Region of Origin	0.029656
Region of Incident	-0.004398
Country of Origin	0.001795
Number of Dead	1.000000
Minimum Estimated Number of Missing	0.208926
Total Number of Dead and Missing	0.641773
Number of Survivors	0.094863
Number of Females	0.147181
Number of Males	0.199589
Number of Children	0.077084
Cause of Death	-0.056879
Migration route	-0.055341
Location of death	-0.042649
Information Source	-0.015144
UNSD Geographical Grouping	0.057422
Latitude	-0.019662
Longitude	0.071367

	Minimum Estimated Number of Missing \
Incident Type	0.070826

Incident year	-0.055558
Reported Month	-0.007721
Region of Origin	0.043659
Region of Incident	-0.043498
Country of Origin	0.002398
Number of Dead	0.208926
Minimum Estimated Number of Missing	1.000000
Total Number of Dead and Missing	0.884053
Number of Survivors	0.131206
Number of Females	0.147635
Number of Males	0.172177
Number of Children	0.453512
Cause of Death	-0.147592
Migration route	-0.092862
Location of death	-0.028315
Information Source	-0.040671
UNSD Geographical Grouping	0.090670
Latitude	0.032465
Longitude	0.054571

	Total Number of Dead and Missing \
Incident Type	0.083497
Incident year	-0.077348
Reported Month	-0.015219
Region of Origin	0.048411
Region of Incident	-0.036213
Country of Origin	0.002738
Number of Dead	0.641773
Minimum Estimated Number of Missing	0.884053
Total Number of Dead and Missing	1.000000
Number of Survivors	0.148230
Number of Females	0.186118
Number of Males	0.230411
Number of Children	0.392485
Cause of Death	-0.142927
Migration route	-0.099272
Location of death	-0.042588
Information Source	-0.039132
UNSD Geographical Grouping	0.098547
Latitude	0.016062
Longitude	0.076903

	Number of Survivors	Number of Females \
Incident Type	0.109269	0.062180
Incident year	-0.044353	-0.015575
Reported Month	-0.006513	0.015927
Region of Origin	0.076922	0.056673

Region of Incident	-0.042863	-0.014861
Country of Origin	0.024192	0.026061
Number of Dead	0.094863	0.147181
Minimum Estimated Number of Missing	0.131206	0.147635
Total Number of Dead and Missing	0.148230	0.186118
Number of Survivors	1.000000	0.029244
Number of Females	0.029244	1.000000
Number of Males	0.059640	0.217927
Number of Children	0.024462	0.120186
Cause of Death	-0.074721	-0.078089
Migration route	-0.121244	0.021933
Location of death	-0.026061	0.001855
Information Source	-0.041555	-0.033250
UNSD Geographical Grouping	0.107444	0.075876
Latitude	0.040866	0.005138
Longitude	0.068794	0.041674

	Number of Males	Number of Children \
Incident Type	0.148464	0.027818
Incident year	0.027849	-0.030011
Reported Month	-0.024489	-0.003678
Region of Origin	0.030489	0.045628
Region of Incident	-0.001823	0.004213
Country of Origin	-0.032380	0.001961
Number of Dead	0.199589	0.077084
Minimum Estimated Number of Missing	0.172177	0.453512
Total Number of Dead and Missing	0.230411	0.392485
Number of Survivors	0.059640	0.024462
Number of Females	0.217927	0.120186
Number of Males	1.000000	0.061169
Number of Children	0.061169	1.000000
Cause of Death	-0.062073	-0.039434
Migration route	-0.049471	-0.011727
Location of death	-0.033049	-0.017800
Information Source	-0.039937	-0.012138
UNSD Geographical Grouping	0.071890	0.043456
Latitude	0.003240	-0.006756
Longitude	0.038784	0.053277

	Cause of Death	Migration route \
Incident Type	-0.099913	-0.055576
Incident year	-0.095503	-0.099958
Reported Month	0.020247	0.006460
Region of Origin	-0.028682	-0.261433
Region of Incident	0.283421	-0.168905
Country of Origin	-0.011881	0.385098
Number of Dead	-0.056879	-0.055341

Minimum Estimated Number of Missing	-0.147592	-0.092862
Total Number of Dead and Missing	-0.142927	-0.099272
Number of Survivors	-0.074721	-0.121244
Number of Females	-0.078089	0.021933
Number of Males	-0.062073	-0.049471
Number of Children	-0.039434	-0.011727
Cause of Death	1.000000	0.133395
Migration route	0.133395	1.000000
Location of death	0.013312	-0.213810
Information Source	0.148667	0.120960
UNSD Geographical Grouping	-0.088946	-0.270369
Latitude	-0.233722	-0.203507
Longitude	0.087025	-0.354839

	Location of death	Information Source \
Incident Type	-0.042855	-0.030788
Incident year	0.029271	-0.159215
Reported Month	-0.012376	-0.016946
Region of Origin	0.072978	-0.187777
Region of Incident	0.045629	0.023146
Country of Origin	-0.110229	0.264236
Number of Dead	-0.042649	-0.015144
Minimum Estimated Number of Missing	-0.028315	-0.040671
Total Number of Dead and Missing	-0.042588	-0.039132
Number of Survivors	-0.026061	-0.041555
Number of Females	0.001855	-0.033250
Number of Males	-0.033049	-0.039937
Number of Children	-0.017800	-0.012138
Cause of Death	0.013312	0.148667
Migration route	-0.213810	0.120960
Location of death	1.000000	0.006075
Information Source	0.006075	1.000000
UNSD Geographical Grouping	-0.061893	-0.151213
Latitude	0.088876	-0.120756
Longitude	0.023146	-0.215984

	UNSD Geographical Grouping	Latitude \
Incident Type	0.050894	0.043708
Incident year	0.052416	0.090723
Reported Month	0.021245	-0.005279
Region of Origin	0.617495	0.172293
Region of Incident	0.628374	-0.060237
Country of Origin	-0.130228	-0.187679
Number of Dead	0.057422	-0.019662
Minimum Estimated Number of Missing	0.090670	0.032465
Total Number of Dead and Missing	0.098547	0.016062
Number of Survivors	0.107444	0.040866

Number of Females	0.075876	0.005138
Number of Males	0.071890	0.003240
Number of Children	0.043456	-0.006756
Cause of Death	-0.088946	-0.233722
Migration route	-0.270369	-0.203507
Location of death	-0.061893	0.088876
Information Source	-0.151213	-0.120756
UNSD Geographical Grouping	1.000000	0.231927
Latitude	0.231927	1.000000
Longitude	0.504979	-0.028584

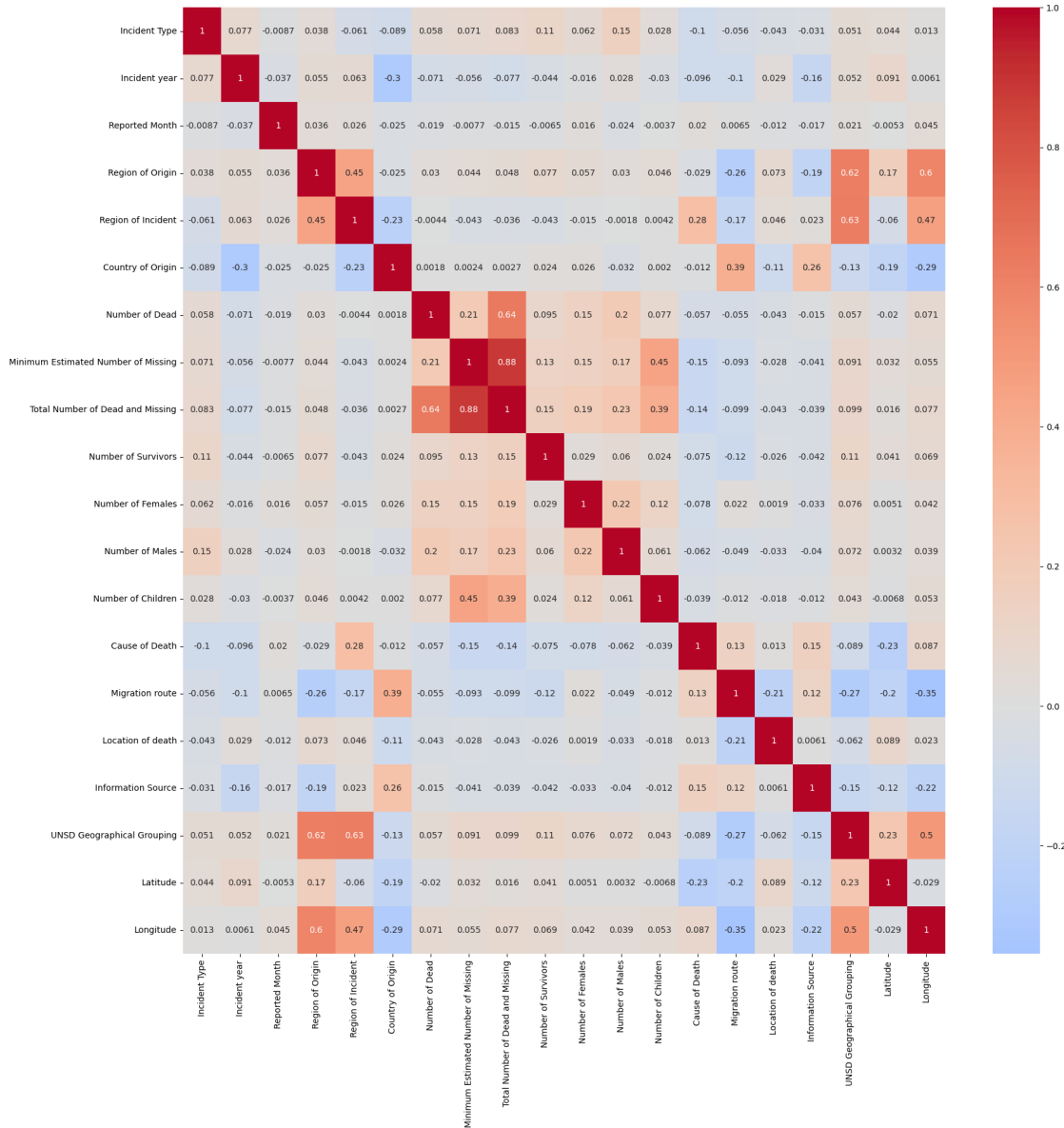
	Longitude
Incident Type	0.013392
Incident year	0.006074
Reported Month	0.044589
Region of Origin	0.599155
Region of Incident	0.469840
Country of Origin	-0.294024
Number of Dead	0.071367
Minimum Estimated Number of Missing	0.054571
Total Number of Dead and Missing	0.076903
Number of Survivors	0.068794
Number of Females	0.041674
Number of Males	0.038784
Number of Children	0.053277
Cause of Death	0.087025
Migration route	-0.354839
Location of death	0.023146
Information Source	-0.215984
UNSD Geographical Grouping	0.504979
Latitude	-0.028584
Longitude	1.000000

```
[ ]: Corr_Matrix = df.corr()

# Set up the figure and plot the heatmap
plt.figure(figsize=(20, 20))
sns.heatmap(Corr_Matrix, annot=True, cmap='coolwarm', center=0)
plt.show()
```

<ipython-input-33-e7c2484fc002>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
Corr_Matrix = df.corr()
```



### Top 5 most positively Correlated

```
[ ]: print('Top 5 Most Positively Correlated to the Total Number of Dead and Missing')
Corr_Matrix['Total Number of Dead and Missing'].sort_values(ascending=False).head(5)
```

Top 5 Most Positively Correlated to the Total Number of Dead and Missing

```
[ ]: Total Number of Dead and Missing      1.000000
      Minimum Estimated Number of Missing    0.884053
```

```
Number of Dead          0.641773
Number of Children      0.392485
Number of Males         0.230411
Name: Total Number of Dead and Missing, dtype: float64
```

### *Top 5 most Negatively Correlated*

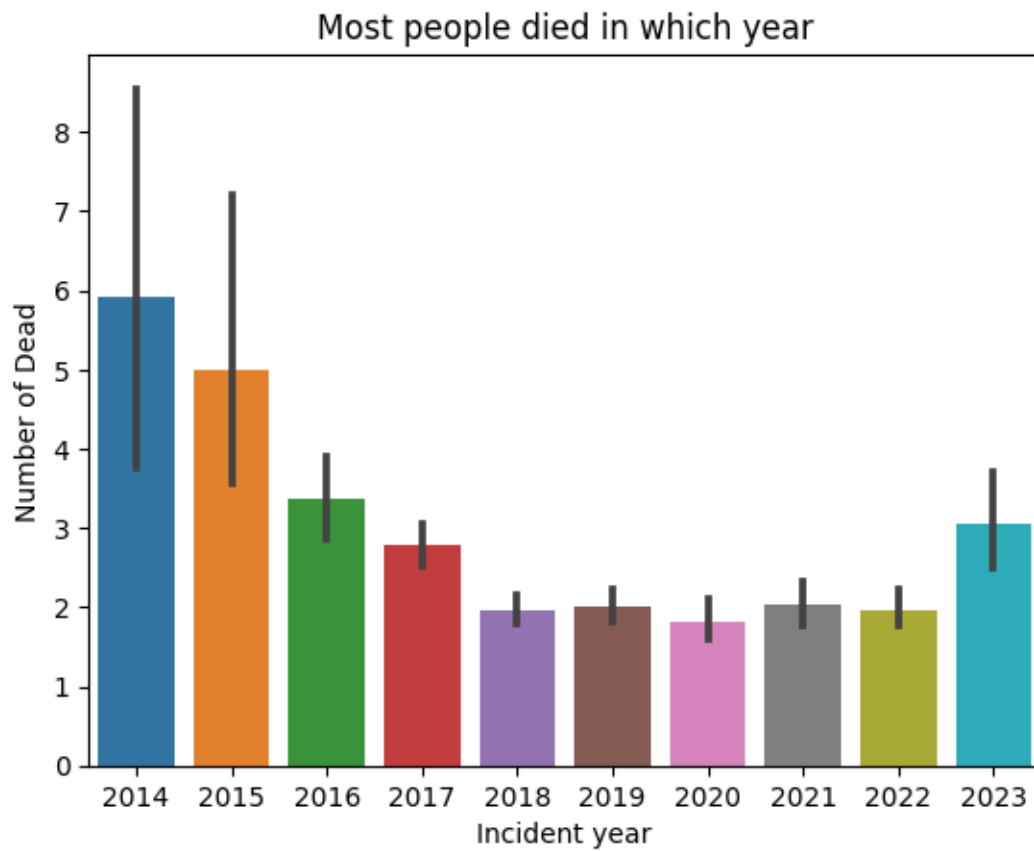
```
[ ]: print('Top 5 Most Negatively Correlated to Total Number of Dead and Missing')
      Corr_Matrix['Total Number of Dead and Missing'].sort_values(ascending=True).
      ↪head(5)
```

Top 5 Most Negatively Correlated to Total Number of Dead and Missing

```
[ ]: Cause of Death      -0.142927
      Migration route    -0.099272
      Incident year      -0.077348
      Location of death  -0.042588
      Information Source  -0.039132
      Name: Total Number of Dead and Missing, dtype: float64
```

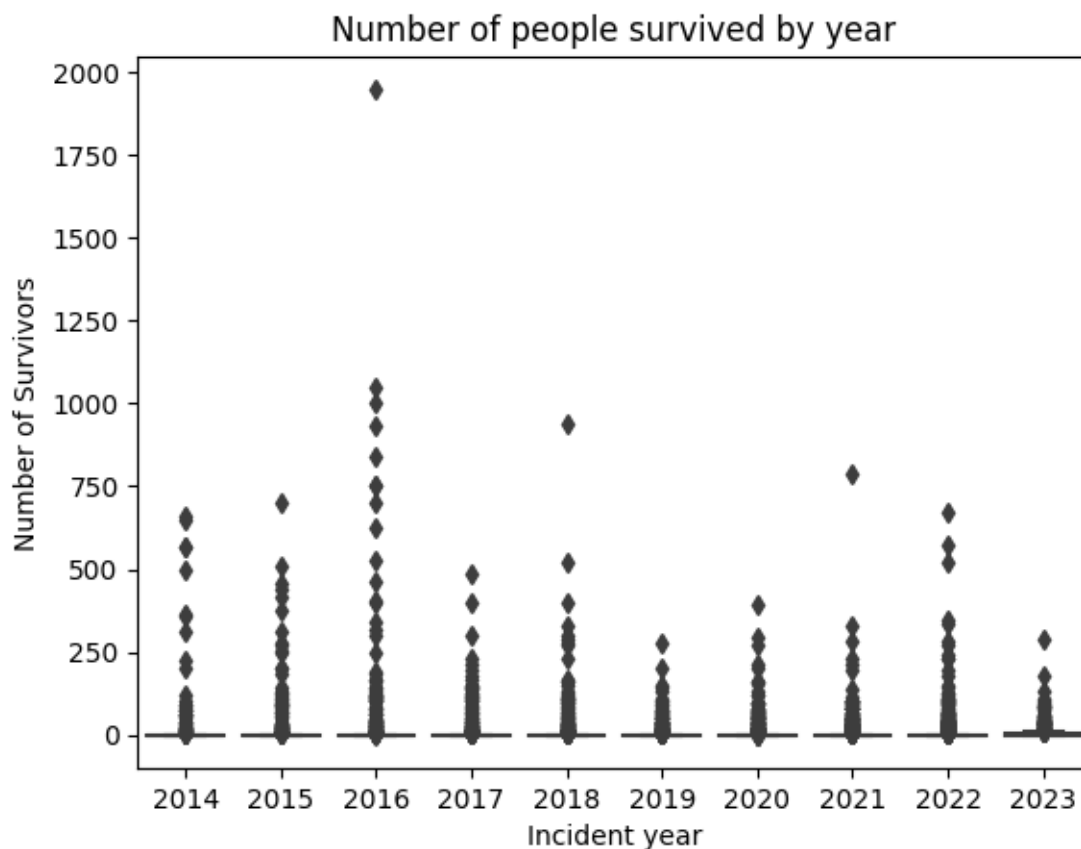
### *Exploratory analysis & Visualization*

```
[ ]: sns.barplot(data= df, x='Incident year',y= 'Number of Dead')
      plt.title('Most people died in which year');
```



```
[ ]: sns.boxplot(data = df, x= 'Incident year', y = 'Number of Survivors')  
plt.title('Number of people survived by year');
```





### *Dropping unwanted columns and rows*

```
[ ]: # dropping Coordinates along with unwanted columns found by correlation matrix
df=df.drop(['Coordinates','Country of Origin'],axis=1)
df
```

```
[ ]:
      Incident Type  Incident year  Reported Month  Region of Origin \
0                1          2014                4                2
1                1          2014                4               10
2                1          2014                4               10
3                1          2014                4                2
4                1          2014                4               13
...              ...            ...              ...              ...
13015             1          2023                5               31
13016             1          2023                5               30
13017             1          2023                5               29
13018             1          2023                5               12
13019             1          2023                5               30
```

```
      Region of Incident  Number of Dead \
```

0	8	1.0
1	8	1.0
2	8	1.0
3	8	1.0
4	5	1.0
...	...	...
13015	15	4.0
13016	15	2.0
13017	9	13.0
13018	9	6.0
13019	14	16.0

	Minimum Estimated Number of Missing	Total Number of Dead and Missing \
0	0	1
1	0	1
2	0	1
3	0	1
4	0	1
...	...	...
13015	0	4
13016	0	2
13017	0	13
13018	0	6
13019	37	53

	Number of Survivors	Number of Females	Number of Males \
0	0	0	1
1	0	0	0
2	0	0	0
3	0	0	1
4	2	0	1
...	...	...	...
13015	0	0	4
13016	0	0	2
13017	6	0	0
13018	48	0	0
13019	2	2	0

	Number of Children	Cause of Death	Migration route	Location of death \
0	0	10	19	4562
1	0	10	19	4562
2	0	10	19	4562
3	0	14	19	7373
4	0	7	21	1362
...	...	...	...	...
13015	0	13	18	2506
13016	0	13	18	678

13017	0	1	23	4281
13018	0	1	23	7056
13019	0	1	23	4165

	Information Source	UNSD Geographical Grouping	Latitude	Longitude
0	2784	8	31.650259	-110.366453
1	2784	8	31.597130	-111.737560
2	2784	8	31.940260	-113.011250
3	2521	8	31.506777	-109.315632
4	890	9	59.155100	28.000000
...	...	...	...	...
13015	277	18	40.912713	26.369657
13016	3127	18	41.716972	26.351489
13017	415	15	23.728361	-15.901632
13018	1114	15	35.171874	-2.903182
13019	1840	15	14.718707	-17.506255

[13020 rows x 19 columns]

### Finding Duplicates

```
[ ]: df.duplicated().sum()
```

```
[ ]: 644
```

```
[ ]: # dropping duplicate rows
df.drop_duplicates(keep='first', inplace=True)
df
```

```
[ ]: Incident Type Incident year Reported Month Region of Origin \
0          1          2014          4          2
1          1          2014          4          10
2          1          2014          4          10
3          1          2014          4          2
4          1          2014          4          13
...
13015      1          2023          5          31
13016      1          2023          5          30
13017      1          2023          5          29
13018      1          2023          5          12
13019      1          2023          5          30
```

	Region of Incident	Number of Dead	\
0	8	1.0	
1	8	1.0	
2	8	1.0	
3	8	1.0	

4	5	1.0
...	...	...
13015	15	4.0
13016	15	2.0
13017	9	13.0
13018	9	6.0
13019	14	16.0

	Minimum Estimated Number of Missing	Total Number of Dead and Missing \
0	0	1
1	0	1
2	0	1
3	0	1
4	0	1
...	...	...
13015	0	4
13016	0	2
13017	0	13
13018	0	6
13019	37	53

	Number of Survivors	Number of Females	Number of Males \
0	0	0	1
1	0	0	0
2	0	0	0
3	0	0	1
4	2	0	1
...	...	...	...
13015	0	0	4
13016	0	0	2
13017	6	0	0
13018	48	0	0
13019	2	2	0

	Number of Children	Cause of Death	Migration route	Location of death \
0	0	10	19	4562
1	0	10	19	4562
2	0	10	19	4562
3	0	14	19	7373
4	0	7	21	1362
...	...	...	...	...
13015	0	13	18	2506
13016	0	13	18	678
13017	0	1	23	4281
13018	0	1	23	7056
13019	0	1	23	4165

	Information Source	UNSD Geographical Grouping	Latitude	Longitude
0	2784	8	31.650259	-110.366453
1	2784	8	31.597130	-111.737560
2	2784	8	31.940260	-113.011250
3	2521	8	31.506777	-109.315632
4	890	9	59.155100	28.000000
...	...	...	...	...
13015	277	18	40.912713	26.369657
13016	3127	18	41.716972	26.351489
13017	415	15	23.728361	-15.901632
13018	1114	15	35.171874	-2.903182
13019	1840	15	14.718707	-17.506255

[12376 rows x 19 columns]

```
[ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 12376 entries, 0 to 13019
```

```
Data columns (total 19 columns):
```

#	Column	Non-Null Count	Dtype
0	Incident Type	12376 non-null	int64
1	Incident year	12376 non-null	int64
2	Reported Month	12376 non-null	int64
3	Region of Origin	12376 non-null	int64
4	Region of Incident	12376 non-null	int64
5	Number of Dead	12376 non-null	float64
6	Minimum Estimated Number of Missing	12376 non-null	int64
7	Total Number of Dead and Missing	12376 non-null	int64
8	Number of Survivors	12376 non-null	int64
9	Number of Females	12376 non-null	int64
10	Number of Males	12376 non-null	int64
11	Number of Children	12376 non-null	int64
12	Cause of Death	12376 non-null	int64
13	Migration route	12376 non-null	int64
14	Location of death	12376 non-null	int64
15	Information Source	12376 non-null	int64
16	UNSD Geographical Grouping	12376 non-null	int64
17	Latitude	12376 non-null	float64
18	Longitude	12376 non-null	float64

```
dtypes: float64(3), int64(16)
```

```
memory usage: 1.9 MB
```

### Importing Libraries

```
[ ]: from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LinearRegression
```

```

from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import r2_score, mean_absolute_error,
↳mean_squared_error, mean_absolute_percentage_error

```

### *Splitting the Dataset*

```

[ ]: x = df.drop(columns=['Total Number of Dead and Missing'])
y = df['Total Number of Dead and Missing']

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2,
↳random_state=42)
# Display the shapes of the resulting datasets
print("X_train shape:", x_train.shape)
print("X_test shape:", x_test.shape)
print("y_train shape:", y_train.shape)
print("y_test shape:", y_test.shape)

```

```

X_train shape: (9900, 18)
X_test shape: (2476, 18)
y_train shape: (9900,)
y_test shape: (2476,)

```

### *Model Building and Analysis*

```

[ ]: models = {
    'Random Forest': RandomForestRegressor(random_state=42),
    'Linear Regression': LinearRegression()
}
best_model = None
best_r2 = 0

for model_name, model in models.items():
    model.fit(x_train, y_train)
    y_pred = model.predict(x_test)

    # Evaluate the model
    r2 = r2_score(y_test, y_pred)
    mape = mean_absolute_percentage_error(y_test, y_pred)
    mae = mean_absolute_error(y_test, y_pred)
    mse = mean_squared_error(y_test, y_pred)
    submit = pd.DataFrame()
    submit['Actual Number of Dead'] = y_test
    submit['Predict Number of Dead'] = y_pred
    submit = submit.reset_index()
    r2 = r2_score(y_test, y_pred)
    if r2 > best_r2:
        best_r2 = r2
        best_model = model_name

```

```

print(f'{model_name}:') # the f-string formatting is used to
↳ embed variables (r2, mape, mae, mse) directly into the strings.
print(f'R2 Score: {r2:.2f}')
print(f'Mean Absolute Percentage Error(MAPE):{mape:.2f}')
print(f'Mean Absolute Error (MAE): {mae:.2f}')
print(f'Mean Squared Error (MSE): {mse:.2f}') # 2f ==>till 2
↳ decimals value will be displayed
print(submit.head(5))

print('-----')
print(f"The best performing model is: {best_model} with accuracy: {best_r2:.
↳ 2f}")

```

Random Forest:

R2 Score: 0.95

Mean Absolute Percentage Error(MAPE):0.00

Mean Absolute Error (MAE): 0.23

Mean Squared Error (MSE): 17.38

	index	Actual Number of Dead	Predict_Number of Dead
0	3794	1	1.00
1	5020	6	6.11
2	7631	1	1.00
3	3234	2	2.00
4	4528	114	114.38

-----

Linear Regression:

R2 Score: 1.00

Mean Absolute Percentage Error(MAPE):0.00

Mean Absolute Error (MAE): 0.00

Mean Squared Error (MSE): 0.00

	index	Actual Number of Dead	Predict_Number of Dead
0	3794	1	1.0
1	5020	6	6.0
2	7631	1	1.0
3	3234	2	2.0
4	4528	114	114.0

-----

The best performing model is: Linear Regression with accuracy: 1.00

## Conclusion

Best Model:

The Linear Regression model appears to outperform the Random Forest model based on the evaluation metrics provided. It achieved perfect predictions on the test data, indicating an exact match between predicted and actual values. The Linear Regression model has an R2 score of 1.00, meaning it explains all the variance in the target variable based on the features.

### Overfitting:

The perfect performance of the Linear Regression model (R2 score of 1.00) on the test set might suggest overfitting, especially when the training and testing datasets are the same. On the other hand, the Random Forest model, although slightly lower in R2 score (0.95), exhibits a reasonable level of performance without perfect accuracy on the test set. This might suggest that it is not overfitting as much as the Linear Regression model.