

PHASE 3: DEVELOPMENT PART 1

ANALYSIS OBJECTIVES:

The objective is to assess and highlight variations in the daily COVID-19 cases and deaths within the European Union and European Economic Area (EU/EEA) member countries. This analysis aims to compare and contrast the mean values to identify regional trends and disparities, while also examining standard deviations to understand the extent of data variability, severity, and trends of COVID-19 in different EU/EEA countries.

PREPROCESSING OF DATASET AND CLEANING THE DATA:

Cleaning a dataset involves the process of preparing data for analysis by identifying and rectifying inconsistencies, errors, and missing values. This procedure is essential for accurate and reliable data-driven insights. Cleaning ensures that the dataset is structured and consistent, making it ready for further analysis and modelling. This crucial step guarantees that data-driven decisions are based on reliable, high-quality information.

Dataset : <https://www.kaggle.com/datasets/chakradharmattapalli/covid-19-cases>

=====**PYTHON CODE (JUPYTER NOTEBOOK) **=====

COVID-19 Cases Analysis

In [1]:

```
# Reading the Excel file into a Pandas DataFrame
import pandas as pd
file_path = r'C:\Users\sankar\Desktop\Covid_19_cases4.xlsx'

# Load the Excel file into a Pandas DataFrame
data = pd.read_excel(file_path)
print(data)
```

	dateRep	day	month	year	cases	deaths	countriesAndTerritories
0	2021-05-31	31	5	2021	366	5	Austria
1	2021-05-30	30	5	2021	570	6	Austria
2	2021-05-29	29	5	2021	538	11	Austria
3	2021-05-28	28	5	2021	639	4	Austria
4	2021-05-27	27	5	2021	405	19	Austria
...
2725	2021-03-06	6	3	2021	3455	17	Sweden
2726	2021-03-05	5	3	2021	4069	12	Sweden
2727	2021-03-04	4	3	2021	4884	14	Sweden
2728	2021-03-03	3	3	2021	4876	19	Sweden
2729	2021-03-02	2	3	2021	6191	19	Sweden

[2730 rows x 7 columns]

In [2]:

```
# Creating copy of original data
cdata=data.copy()
```

In [3]:

```
# Structure of the dataset
cdata.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2730 entries, 0 to 2729
Data columns (total 7 columns):
 #   Column                                Non-Null Count  Dtype  
---  -
 0   dateRep                              2730 non-null   datetime64[ns]
 1   day                                  2730 non-null   int64   
 2   month                               2730 non-null   int64   
 3   year                                 2730 non-null   int64   
 4   cases                               2730 non-null   int64   
 5   deaths                              2730 non-null   int64   
 6   countriesAndTerritories             2730 non-null   object  
dtypes: datetime64[ns](1), int64(5), object(1)
memory usage: 149.4+ KB
```

In [4]:

```
# Summary of numerical variables
summary_num = cdata.describe()
print(summary_num)
```

	day	month	year	cases	deaths
count	2730.000	2730.000	2730.000	2730.000	2730.000
mean	16.000	4.011	2021.000	3661.011	65.292
std	8.766	0.819	0.000	6490.510	113.957
min	1.000	3.000	2021.000	-2001.000	-3.000
25%	8.000	3.000	2021.000	361.250	2.000
50%	16.000	4.000	2021.000	926.500	14.500
75%	24.000	5.000	2021.000	3916.250	72.000
max	31.000	5.000	2021.000	53843.000	956.000

In [5]:

```
#Summary of categorical variables
summary_cate = cdata.describe(include = "O")
print(summary_cate)
```

	countriesAndTerritories
count	2730
unique	30
top	Austria
freq	91

In [6]:

```
# Removing duplicate records
cdata.drop_duplicates(keep='first',inplace=True)
```

In [7]:

```
# Check for missing values
cdata.isnull()
print('Data columns with null values:\n', cdata.isnull().sum())
```

```
Data columns with null values:
  dateRep          0
  day            0
  month          0
  year          0
  cases          0
  deaths         0
  countriesAndTerritories  0
dtype: int64
```

In [8]:

```
# Calculate Mean Daily Cases
mean_daily_cases = cdata['cases'].mean()
print("Mean Daily Cases:", mean_daily_cases)

# Calculate Mean Daily Deaths
mean_daily_deaths = cdata['deaths'].mean()
print("Mean Daily Deaths:", mean_daily_deaths)

# Calculate Standard Deviation of Daily Cases
std_daily_cases = cdata['cases'].std()
print("Standard Deviation of Daily Cases:", std_daily_cases)

# Calculate Standard Deviation of Daily Deaths
std_daily_deaths = cdata['deaths'].std()
print("Standard Deviation of Daily Deaths:", std_daily_deaths)

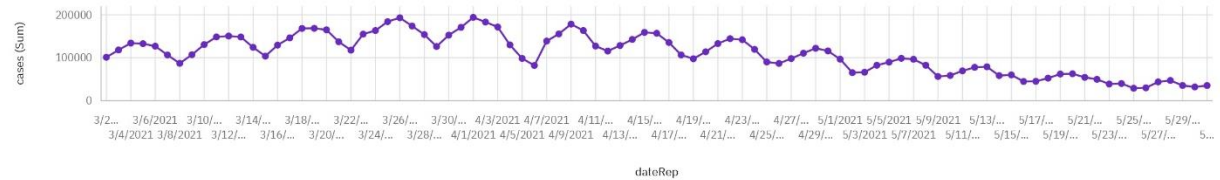
Mean Daily Cases: 3661.010989010989
Mean Daily Deaths: 65.29194139194139
Standard Deviation of Daily Cases: 6490.510073102111
Standard Deviation of Daily Deaths: 113.95663405806982
```

```
=====**END OF THE CODE**=====
```

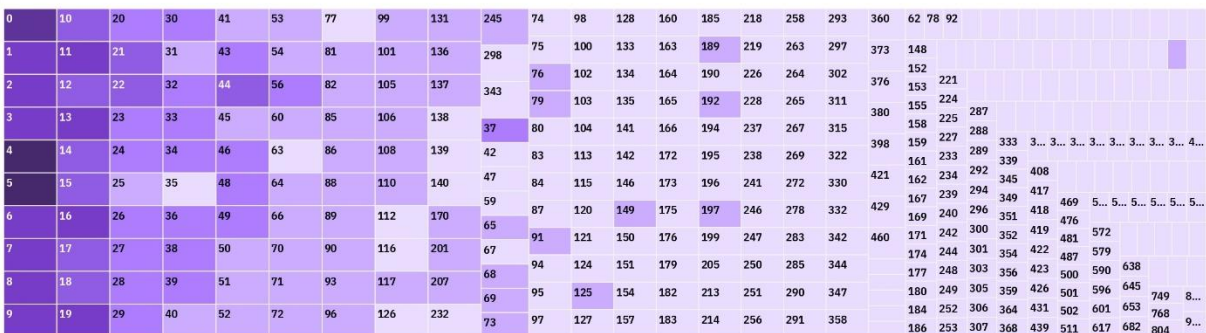
OVERVIEW OF VISUALIZATION OF COVID-19 CASES AND DEATHS IN IBM COGNOS:

ANALYSIS OF CASES:

cases by dateRep



deaths hierarchy colored by countriesAndTerritories and sized by month

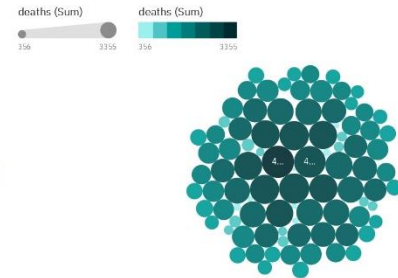


ANALYSIS OF CASES AND DEATHS:

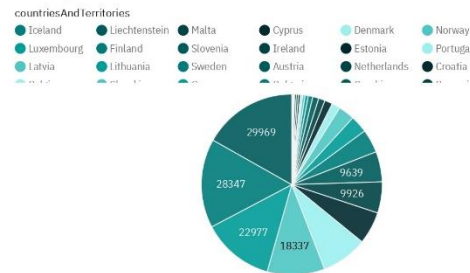
cases and deaths by dateRep



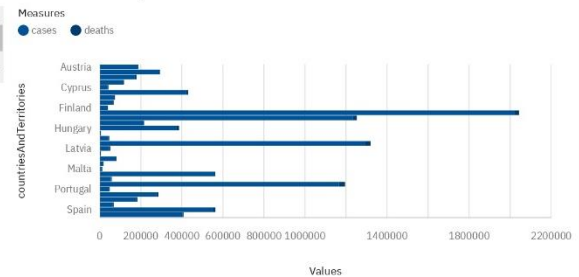
dateRep colored by deaths sized by deaths



deaths by countriesAndTerritories

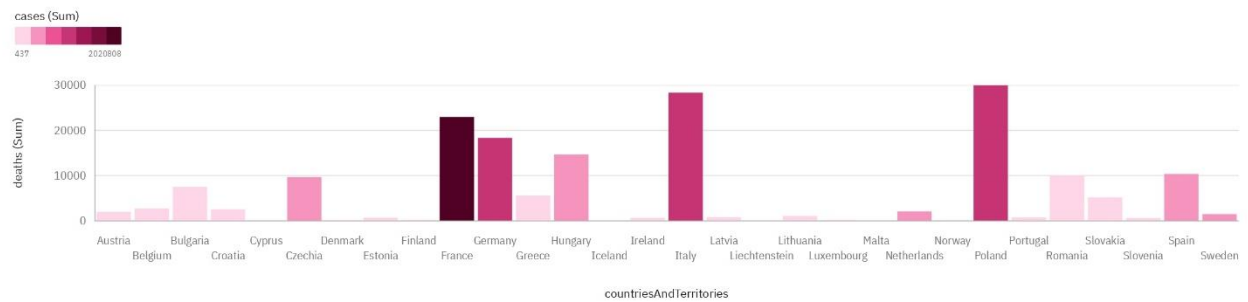


cases and deaths by countriesAndTerritories

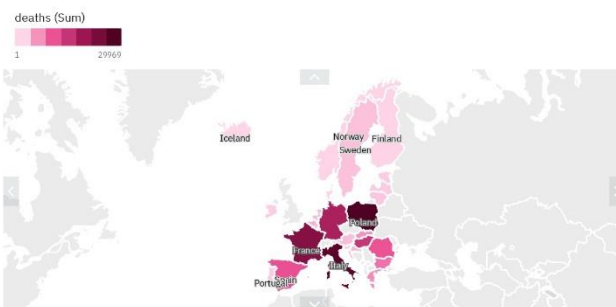


ANALYSIS OF DEATHS:

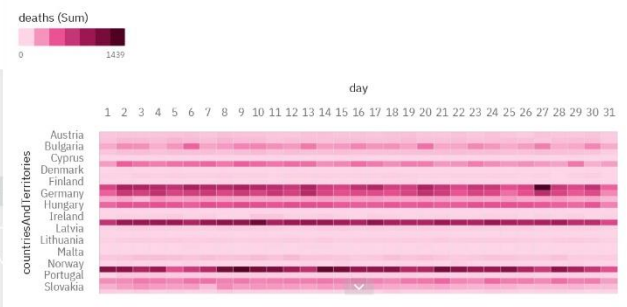
deaths by countriesAndTerritories colored by cases



countriesAndTerritories, deaths



deaths by countriesAndTerritories and day



CONCLUSION:

In this phase, we initiated the development of our COVID-19 cases analysis project. We outlined our objectives, which involve leveraging IBM Cognos for visualization, providing a powerful platform for data exploration and presentation. The initial focus was on data preprocessing and cleaning to guarantee the data's accuracy and reliability. Also, we had an overview on visualization of the cases and deaths using IBM Cognos with various types of visualizations charts.