

Covid-19-data

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The data and data sources

- Confirmed cases and deaths: this data is collected from the World Health
 Organization Coronavirus Dashboard. The cases & deaths dataset is updated
 daily.
 - ➤ Note 1: Time/date stamps reflect when the data was last updated by WHO. Due to the time required to process and validate the incoming data, there is a delay between reporting to WHO and the update of the dashboard.
 - Note 2: Counts and corrections made after these times will be carried forward to the next reporting cycle for that specific region. Delayed reporting for any specific country, territory or area may result in pooled counts for multiple days being presented, with a retrospective update to counts on previous days to accurately reflect trends. Significant data errors detected or reported to WHO may be corrected at more frequent intervals.
- Hospitalizations and intensive care unit (ICU) admissions: our data is collected from official sources and collated by Our World in Data. The complete list of country-by-country sources is available here.
- Testing for COVID-19: this data is collected by the Our World in Data team from official reports; you can find further details in our post on COVID-19 testing, including our checklist of questions to understand testing data, information on geographical and temporal coverage, and detailed country-by-country source information. On 23 June 2022, we stopped adding new datapoints to our COVID-19 testing dataset. You can read more here.
- Vaccinations against COVID-19: this data is collected by the *Our World in Data* team from official reports.
- Other variables: this data is collected from a variety of sources (United Nations, World Bank, Global Burden of Disease, Blavatnik School of Government, etc.).
 More information is available in <u>our codebook</u>.

THE DATASET HAS THE FOLLOWING ATTRIBUTES:

1. iso code:

 Description: ISO 3166-1 alpha-3 – three-letter country codes. Note that OWID-defined regions (e.g. continents like 'Europe') contain prefix 'OWID '.

2. continent:

• **Description:** Continent of the geographical location.

3. location:

Description: Geographical location.

4. date:

• **Description:** Date of observation.

5. total cases:

• **Description:** Total confirmed cases of COVID-19. Counts can include probable cases, where reported.

6. new cases:

 Description: New confirmed cases of COVID-19. Counts can include probable cases, where reported. In rare cases where our source reports a negative daily change due to a data correction, we set this metric to NA.

7. new_cases_smoothed:

Description: New confirmed cases of COVID-19 (7-day smoothed).
 Counts can include probable cases, where reported.

8. total_deaths:

• **Description:** Total deaths attributed to COVID-19. Counts can include probable deaths, where reported.

9. new_deaths:

 Description: New deaths attributed to COVID-19. Counts can include probable deaths, where reported. In rare cases where our source reports a negative daily change due to a data correction, we set this metric to NA.

10. new deaths smoothed:

• **Description:** New deaths attributed to COVID-19 (7-day smoothed). Counts can include probable deaths, where reported.

11. total_cases_per_million:

Description: Total confirmed cases of COVID-19 per 1,000,000 people.
 Counts can include probable cases, where reported.

12. new_cases_per_million:

• **Description:** New confirmed cases of COVID-19 per 1,000,000 people. Counts can include probable cases, where reported.

13. new_cases_smoothed_per_million:

• **Description:** New confirmed cases of COVID-19 (7-day smoothed) per 1,000,000 people. Counts can include probable cases, where reported.

14. total_deaths_per_million:

• **Description:** Total deaths attributed to COVID-19 per 1,000,000 people. Counts can include probable deaths, where reported.

15. new_deaths_per_million:

Description: New deaths attributed to COVID-19 per 1,000,000 people.
 Counts can include probable deaths, where reported.

16. new_deaths_smoothed_per_million:

• **Description:** New deaths attributed to COVID-19 (7-day smoothed) per 1,000,000 people. Counts can include probable deaths, where reported.

17. reproduction_rate:

 Description: Real-time estimate of the effective reproduction rate (R) of COVID-19. See https://github.com/crondonm/TrackingR/tree/main/Estimates-Database

18. icu patients:

• **Description:** Number of COVID-19 patients in intensive care units (ICUs) on a given day.

19. icu patients per million:

• **Description:** Number of COVID-19 patients in intensive care units (ICUs) on a given day per 1,000,000 people.

20. hosp_patients:

• **Description:** Number of COVID-19 patients in hospital on a given day.

21. hosp_patients_per_million:

• **Description:** Number of COVID-19 patients in hospital on a given day per 1,000,000 people.

22. weekly_icu_admissions:

 Description: Number of COVID-19 patients newly admitted to intensive care units (ICUs) in a given week (reporting date and the preceding 6 days).

23. weekly icu admissions per million:

• **Description:** Number of COVID-19 patients newly admitted to intensive care units (ICUs) in a given week per 1,000,000 people (reporting date and the preceding 6 days).

24. weekly_hosp_admissions:

• **Description:** Number of COVID-19 patients newly admitted to hospitals in a given week (reporting date and the preceding 6 days).

25. weekly_hosp_admissions_per_million:

• **Description:** Number of COVID-19 patients newly admitted to hospitals in a given week per 1,000,000 people (reporting date and the preceding 6 days).

26. total_tests:

• **Description:** Total tests for COVID-19.

27. new_tests:

• **Description:** New tests for COVID-19 (only calculated for consecutive days).

28. total_tests_per_thousand:

• **Description:** Total tests for COVID-19 per 1,000 people.

29. new tests per thousand:

Description: New tests for COVID-19 per 1,000 people.

30. new_tests_smoothed:

• **Description:** New tests for COVID-19 (7-day smoothed). For countries that don't report testing data on a daily basis, we assume that testing changed equally on a daily basis over any periods in which no data was reported. This produces a complete series of daily figures, which is then averaged over a rolling 7-day window.

31. new_tests_smoothed_per_thousand:

Description: New tests for COVID-19 (7-day smoothed) per 1,000 people.

32. positive_rate:

• **Description:** The share of COVID-19 tests that are positive, given as a rolling 7-day average (this is the inverse of tests_per_case).

33. tests_per_case:

• **Description:** Tests conducted per new confirmed case of COVID-19, given as a rolling 7-day average (this is the inverse of positive_rate).

34. tests_units:

Description: Units used by the location to report its testing data. A
country file can't contain mixed units. All metrics concerning testing data
use the specified test unit. Valid units are 'people tested' (number of
people tested), 'tests performed' (number of tests performed. a single
person can be tested more than once in a given day) and 'samples tested'
(number of samples tested. In some cases, more than one sample may
be required to perform a given test.)

35. total vaccinations:

Description: Total number of COVID-19 vaccination doses administered.

36. people_vaccinated:

 Description: Total number of people who received at least one vaccine dose.

37. people fully vaccinated:

• **Description:** Total number of people who received all doses prescribed by the initial vaccination protocol.

38. total boosters:

 Description: Total number of COVID-19 vaccination booster doses administered (doses administered beyond the number prescribed by the vaccination protocol).

39. new vaccinations:

• **Description:** New COVID-19 vaccination doses administered (only calculated for consecutive days).

40. new_vaccinations_smoothed:

Description: New COVID-19 vaccination doses administered (7-day smoothed). For countries that don't report vaccination data on a daily basis, we assume that vaccination changed equally on a daily basis over any periods in which no data was reported. This produces a complete series of daily figures, which is then averaged over a rolling 7-day window.

41. total vaccinations per hundred:

• **Description:** Total number of COVID-19 vaccination doses administered per 100 people in the total population.

42. people vaccinated per hundred:

• **Description:** Total number of people who received at least one vaccine dose per 100 people in the total population.

43. people_fully_vaccinated_per_hundred:

• **Description:** Total number of people who received all doses prescribed by the initial vaccination protocol per 100 people in the total population.

44. total_boosters_per_hundred:

• **Description:** Total number of COVID-19 vaccination booster doses administered per 100 people in the total population.

45. new vaccinations smoothed per million:

• **Description:** New COVID-19 vaccination doses administered (7-day smoothed) per 1,000,000 people in the total population.

46. new_people_vaccinated_smoothed:

• **Description:** Daily number of people receiving their first vaccine dose (7-day smoothed).

47. new_people_vaccinated_smoothed_per_hundred:

• **Description:** Daily number of people receiving their first vaccine dose (7-day smoothed) per 100 people in the total population.

48. stringency_index:

 Description: Government Response Stringency Index: composite measure based on 9 response indicators including school closures, workplace closures, and travel bans, rescaled to a value from 0 to 100 (100 = strictest response).

49. population:

Description: Population (latest available values). See
 https://github.com/owid/covid-19-
 data/blob/master/scripts/input/un/population_latest.csv for a full list of sources.

50. population_density:

• **Description:** Number of people divided by land area, measured in square kilometers, most recent year available.

51. median_age:

• **Description:** Median age of the population, UN projection for 2020.

52. aged_65_older:

• **Description:** Share of the population that is 65 years and older, most recent year available.

53. **aged_70_older:**

Description: Share of the population that is 70 years and older in 2015.

54. gdp_per_capita:

• **Description:** Gross domestic product at purchasing power parity (constant 2011 international dollars), most recent year available.

55. extreme_poverty:

• **Description:** Share of the population living in extreme poverty, most recent year available since 2010.

56. cardiovasc death rate:

• **Description:** Death rate from cardiovascular disease in 2017 (annual number of deaths per 100,000 people).

57. diabetes_prevalence:

• **Description**: Diabetes prevalence (% of population aged 20 to 79) in 2017.

58. female_smokers:

• **Description:** Share of women who smoke, most recent year available.

59. male smokers:

Description: Share of men who smoke, most recent year available.

60. handwashing_facilities:

 Description: Share of the population with basic handwashing facilities on premises, most recent year available.

61. hospital beds per thousand:

• **Description**: Hospital beds per 1,000 people, most recent year available since 2010.

62. life_expectancy:

Description: Life expectancy at birth in 2019.

63. human development index:

 Description: A composite index measuring average achievement in three basic dimensions of human development—a long and healthy life, knowledge and a decent standard of living. Values for 2019, imported from http://hdr.undp.org/en/indicators/137506.

64. excess_mortality:

 Description: Percentage difference between the reported number of weekly or monthly deaths in 2020–2021 and the projected number of deaths for the same period based on previous years. For more information, see https://github.com/owid/covid-19-data/tree/master/public/data/excess mortality.

65. excess_mortality_cumulative:

 Description: Percentage difference between the cumulative number of deaths since 1 January 2020 and the cumulative projected deaths for the same period based on previous years. For more information, see https://github.com/owid/covid-19-data/tree/master/public/data/excess mortality.

66. excess_mortality_cumulative_absolute:

 Description: Cumulative difference between the reported number of deaths since 1 January 2020 and the projected number of deaths for the same period based on previous years. For more information, see https://github.com/owid/covid-19-data/tree/master/public/data/excess_mortality.

67. excess mortality cumulative per million:

• **Description:** Cumulative difference between the reported number of deaths since 1 January 2020 and the projected number of deaths for the same period based on previous years, per million people.

EDA

Data Exploration and Cleaning Report

Initial Data Inspection:

- The dataset contains 355,434 entries and 66 columns.
- The 'iso_code' column was dropped as it was deemed unnecessary.
- the latest date in the dataset: 2023-11-09
- North Korea's total cases on the last day: nan

It seems that the value for North Korea's total cases on the last day is NaN (Not a Number), which typically indicates missing or unavailable data. In many cases, countries may not provide accurate or up-to-date information.

investigating why the data is missing. It could be due to North Korea not reporting COVID-19 cases, lack of data availability, or reporting delays those countries with closed or restricted information policies, may be limited.

• Filter the data for countries with NaN total cases on the last date

```
['North Korea' 'Turkmenistan']
```

droped rows for 'North Korea' and 'Turkmenistan'

Missing Values Analysis:

- Checked for missing values in each column
- Several columns have missing values, ranging from 4.75% to 96.56%. Columns like 'reproduction_rate,' 'icu_patients,' and 'stringency_index' have high percentages of missing values.

continent	16884
location	0
date	0
total_cases	35164
new_cases	9579
new_cases_smoothed	10828
total_deaths	56883
new_deaths	9527
new_deaths_smoothed	10747
total_cases_per_million	35164
new_cases_per_million	9579
new_cases_smoothed_per_million	10828
total_deaths_per_million	56883

new_deaths_per_million	9527
new_deaths_smoothed_per_million	10747
reproduction_rate	167803
icu_patients	314889
icu_patients_per_million	314889
hosp_patients	313542
hosp_patients_per_million	313542
weekly_icu_admissions	342363
<pre>weekly_icu_admissions_per_million</pre>	342363
weekly_hosp_admissions	329238
<pre>weekly_hosp_admissions_per_million</pre>	329238
total_tests	273297
new_tests	277219
total_tests_per_thousand	273297
new_tests_per_thousand	277219
new_tests_smoothed	249377
new_tests_smoothed_per_thousand	249377
positive rate	256693
tests per case	258272
tests units	246561
total_vaccinations	272939
people vaccinated	276356
people fully vaccinated	279677
total boosters	304707
new vaccinations	286936
new vaccinations smoothed	170926
total vaccinations per hundred	272939
people vaccinated per hundred	276356
people fully vaccinated per hundred	279677
total_boosters_per_hundred	304707
new vaccinations smoothed per million	170926
new people vaccinated smoothed	171158
new people vaccinated smoothed per hundred	171158
stringency index	156063
population density	53635
median age	74798
aged 65 older	84589
aged 70 older	77612
gdp per capita	78961
extreme poverty	175334
cardiovasc death rate	79717
diabetes prevalence	65686
female smokers	145787
male smokers	148601
handwashing facilities	218953
hospital beds per thousand	112019
life expectancy	28367
human development index	86791
population	00731
excess mortality cumulative absolute	340409
excess mortality cumulative	340409
excess mortality	340409
excess mortality cumulative per million	340409
	5 10 10 5

Handling Missing Data:

Continent

- Find unique names of countries with missing values in 'continent'
- ['Africa' 'Asia' 'Europe' 'European Union' 'High income' 'Low income'
- 'Lower middle income' 'North America' 'Oceania' 'South America'
- 'Upper middle income' 'World']

using the loc accessor to locate the rows where the 'location' column matches the specified countries and fills the 'continent' column with the value "iso_continent" for those rows.

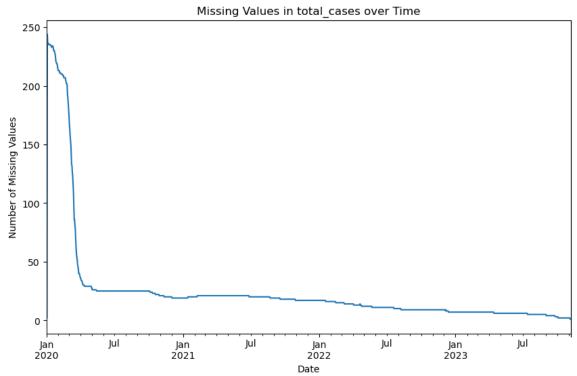
After running the code, checking the specific rows for the countries filled to ensure that the 'continent' column has been updated as expected.

continent

0.000000

location 0.000000 date 0.000000

total_case



Created a pivot table to count the occurrences of missing 'total_cases' for each combination of 'location' and 'date'

0 Afghanista	an 2020-01-03
1 Afghanista	an 2020-01-04 1
2 Afghanista	an 2020-01-05
3 Afghanista	an 2020-01-06 1
4 Afghanista	an 2020-01-07 1
5 Afghanista	an 2020-01-08 1
6 Afghanista	an 2020-01-09 1
7 Afghanista	an 2020-01-10 1
8 Afghanista	an 2020-01-11 1
9 Afghanista	an 2020-01-12 1

Now Replaced missing values in 'total_cases' with zeros.

```
Number of missing values in 'total cases' after replacement: 0
```

more locations to drop

Create a pivot table to count the occurrences of missing 'total_cases' for each combination of 'location' and 'date'

it seems like wales have missing values.

```
Total cases in Wales on the last reported day: 0.0
```

now I found some more countries that have bad report and i want to find those counries that have 0 total_case on their last day of data of that specific location has been report

```
Countries with 0 total cases on their last reported day: ['England' 'Hong Kong' 'Macao' 'Northern Cyprus' 'Northern Ireland' 'Scotland' 'Taiwan' 'Wales' 'Western Sahara']

Number of missing values in 'new_cases' after replacement: 0

Number of missing values in 'new cases smoothed'after replacement: 0
```

new_deaths

Created a pivot table to count the occurrences of missing 'total_cases' for each combination of 'location' and 'date'

it seems like USA have that condition.

```
New deaths in the United States on 2022-03-10: 0.0
```

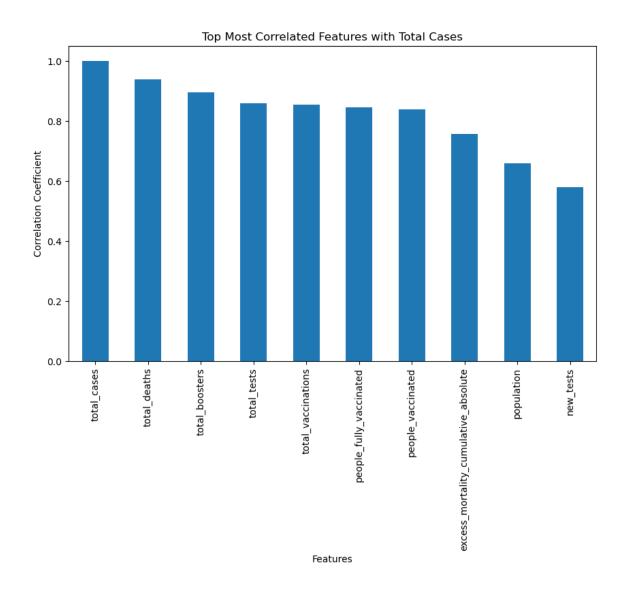
Filled missing values of total_deaths and new_deaths with 0

Now I have these features with zero missing values:

continent	0.000000
location	0.000000
date	0.000000
total cases	0.000000
new_cases	0.000000
new_cases_smoothed	0.000000
total deaths	0.000000
new_deaths	0.000000
new deaths smoothed	0.000000

[♣] Imputed missing values or handled them appropriately based on the nature of the data.

Correlation

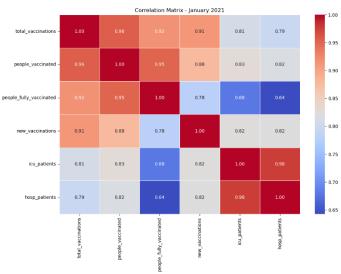


Selecting the correlation values for 'total_cases' and sorting them

total_cases	1.000000
total_deaths	0.938820
total_boosters	0.895218
total_tests	0.859420
total_vaccinations	0.854184
people_fully_vaccinated	0.844591
people_vaccinated	0.838974
excess_mortality_cumulative_absolute	0.755098
population	0.658435
new_tests	0.578069

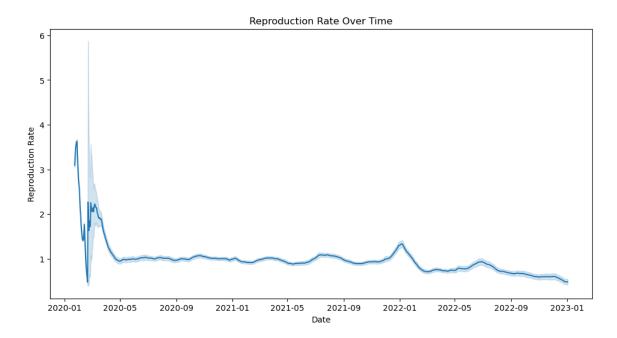
These values represent the strength and direction of the linear relationship between 'total_cases' and each of the listed variables.



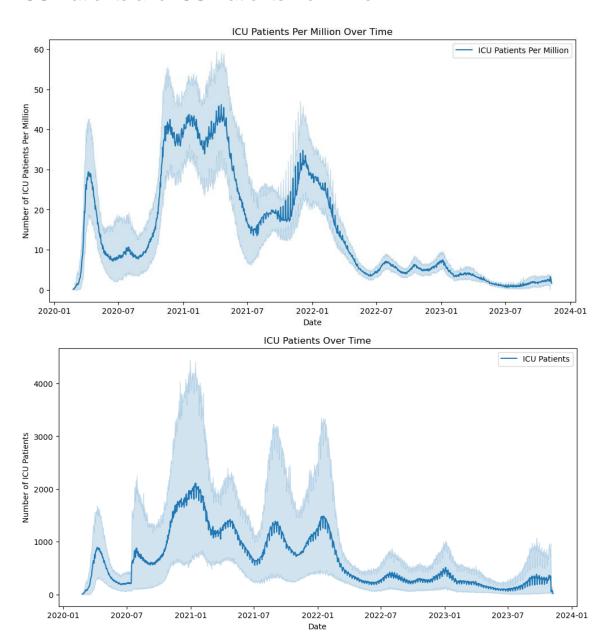


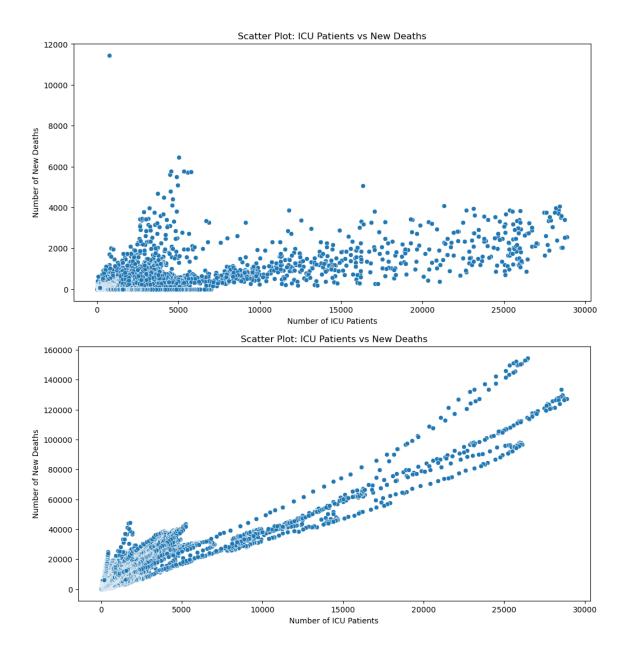
Reproduction Rate:

Line chart over time: Show how the reproduction rate changes over time. Insights: Identify periods of high and low reproduction rates.



ICU Patients and ICU Patients Per Million:





Extracted correlation values related to 'icu_patients'

1.000000
0.941822
0.941201
0.932974
0.903094
0.735466

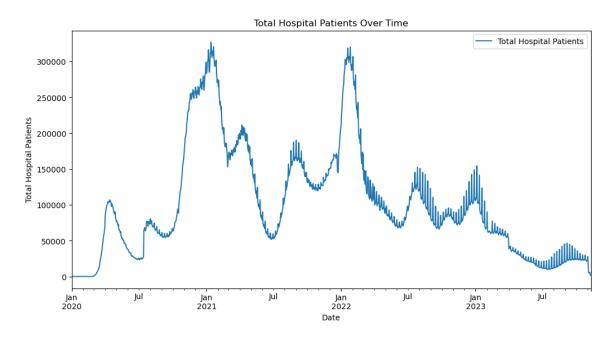
Summary statistics for ICU patients and total cases

```
ICU Patients Summary:
count 33440.000000
mean
          739.720156
std
        2294.376172
             0.000000
min
25%
            25.000000
         112.000000
492.000000
50%
75%
max 28891.000000
Name: icu patients, dtype: float64
Total Cases Summary:
count 3.433120e+05
mean
        6.285071e+06
        3.969616e+07
std
0.000000e+00
25% 4.312000e+03
50% 4.996300
75%
        6.385288e+05
max
        7.718202e+08
Name: total cases, dtype: float64
```

- The average number of ICU patients is approximately 740. The standard deviation is relatively high (2294), indicating significant variability in ICU occupancy. The minimum value is 0, suggesting instances where there were no ICU patients reported. The maximum value is 28,891, indicating a wide range in the number of ICU patients across the dataset. Total Cases:
- The average total number of cases is around 6.29 million. The standard deviation is substantial (about 39.7 million), indicating a large variation in total cases globally. The minimum value is 0, which may represent missing or erroneous data for certain locations. The maximum value is approximately 771.82 million, indicating a broad range in the total number of cases across different regions.

Hospital Patients and Hospital Patients Per Million:

Group the data by date and calculate the total hospital patients for each date. Plot a line chart with the date on the x-axis and the total hospital patients on the y-axis.



- during March 2022 and January 2021 could be indicative of significant events or factors influencing healthcare systems.
- Investigate if there were specific events or factors during March 2022 and January 2021 that could explain the spikes. This could include the emergence of new variants, changes in public health policies, or peaks in infection rates.
- Examine the vaccination rates during these periods. Higher vaccination rates may influence the severity of cases and contribute to a potential decrease in hospitalizations. Public Health Interventions:
- Review any implemented public health interventions or restrictions during these times. Changes in mitigation measures can impact the spread of the virus and subsequent hospitalization rates

Dropping these features

After this analysis I decided to drop these features:

FEATURES_TO_DROP:

- 'reproduction_rate',
- 'icu_patients',
- 'icu_patients_per_million',
- 'hosp_patients',
- 'hosp_patients_per_million',
- 'weekly_icu_admissions',
- 'weekly_icu_admissions_per_million',
- 'weekly_hosp_admissions' 'weekly_hosp_admissions_per_million'

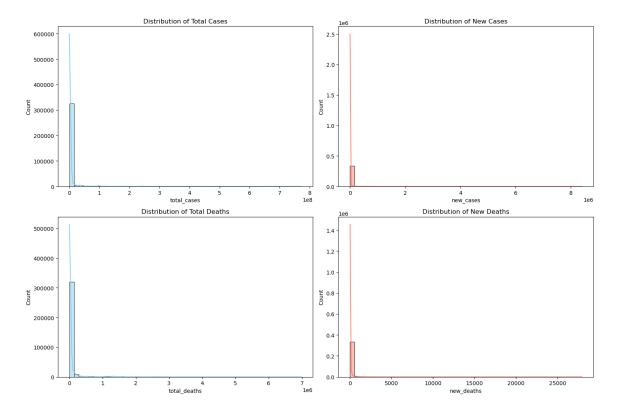
some of features that have many nan values, in the farther analysis has been dropped and do our work on data has remained.

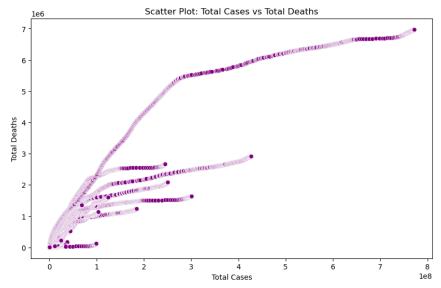
Since after the analysis I realized that null values are meaningful, the graphs drawn will be based on this criterion:

- INCLUDE NULLS
- NO NULLS

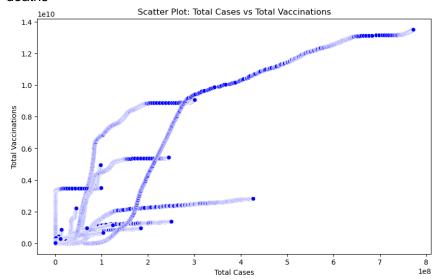
2. Data visualization

visualizing the distribution of 'total_cases', 'new_cases', 'total_deaths', and 'new_deaths'. This will give us an idea of the spread of these variables



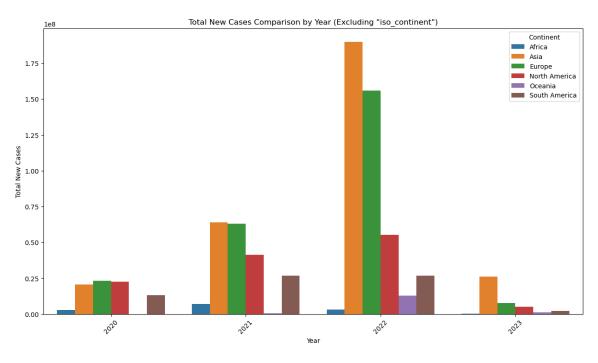


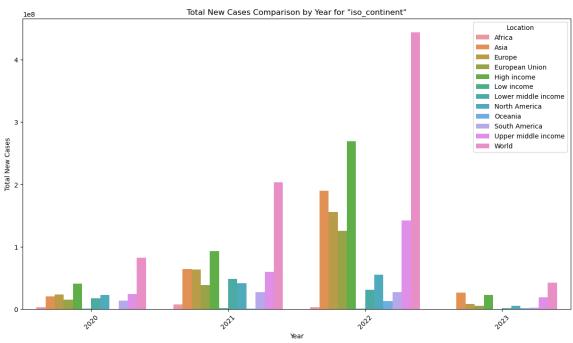
Indeed, there seems to be a positive correlation between 'total_cases' and 'total_deaths,' which is expected—higher total cases often correlate with higher total deaths

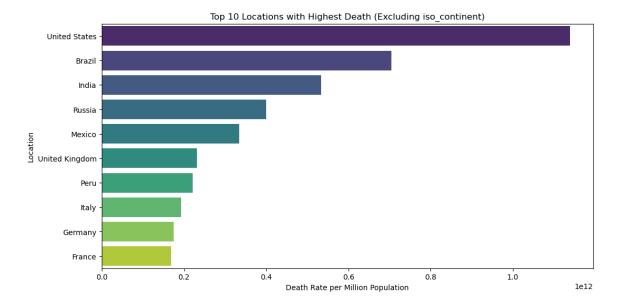


Indeed, a linear relationship suggests that as the total number of cases increases, the total number of vaccinations also tends to increase. This could be an indication of efforts to vaccinate the population in response to the rising number of cases.

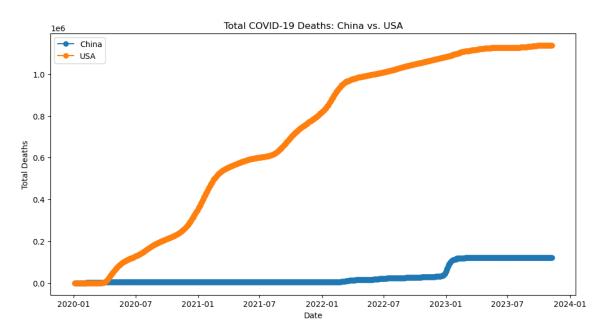
Total New Cases



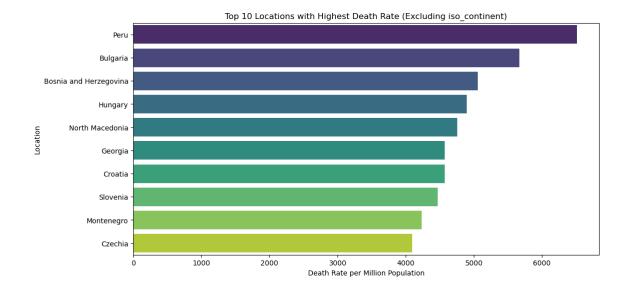




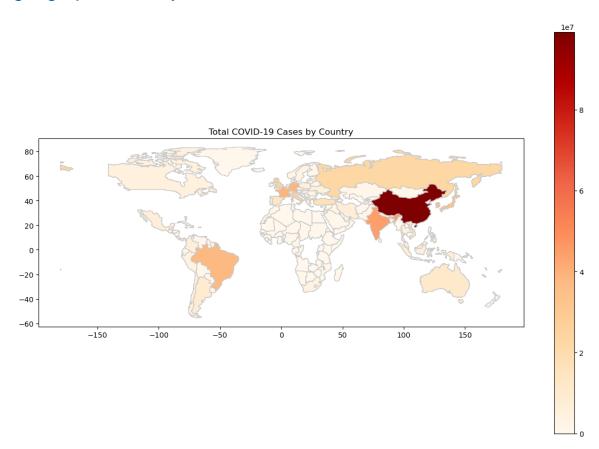
After this chart, I doubted China's statistics and found this information

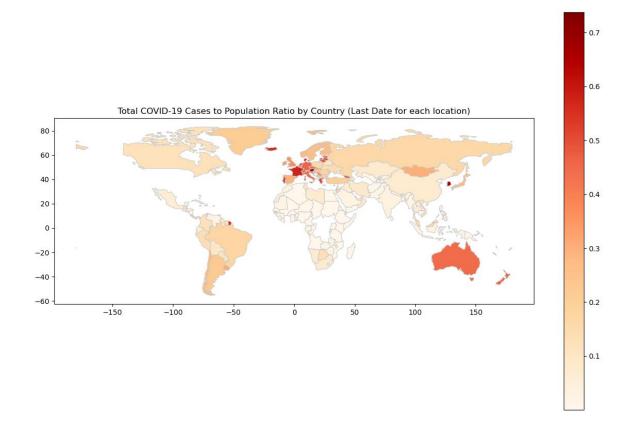


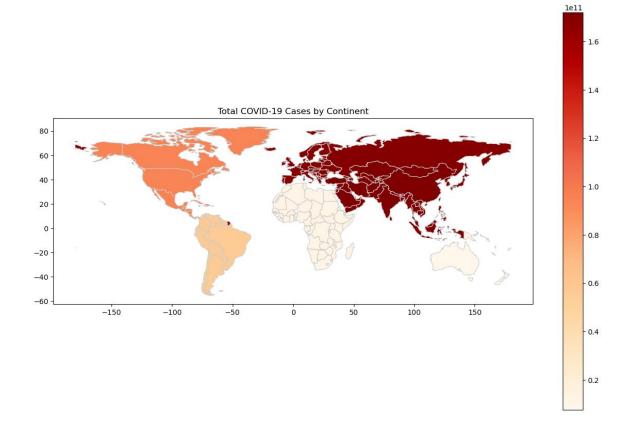
Total deaths in China: 121790.0



In addition to these approaches, also used map to explore geographical analysis on Covid-19.

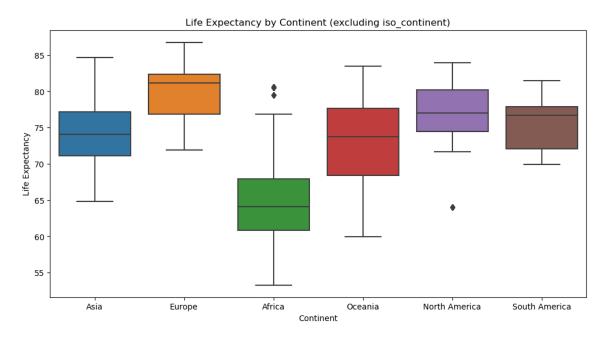




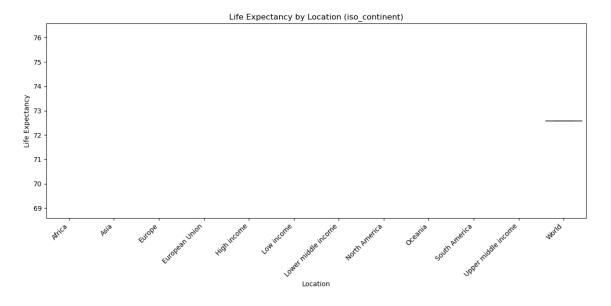


life_expectancy

location	
Monaco 80	6.75
San Marino 84	4.97
Japan 84	4.63
Cayman Islands 83	3.92
Switzerland 83	3.78
Andorra 83	3.73
Singapore 83	3.62
Spain 83	3.56
Italy 83	3.51
Chad 54 Lesotho 54 Nigeria 55 Sierra Leone 55 Somalia 55 Cote d'Ivoire 55	3.28 4.24 4.33 4.69 4.70 7.40 7.78 7.85



	continent	life expectancy
0	Africa	64.10
1	Asia	74.05
2	Europe	81.15
3	North America	77.02
4	Oceani <i>a</i>	73.70
5	South America	76.67



It seems data is not collected correctly for "iso_continent"

Iran's life expectancy: 76.68

Age over 65:

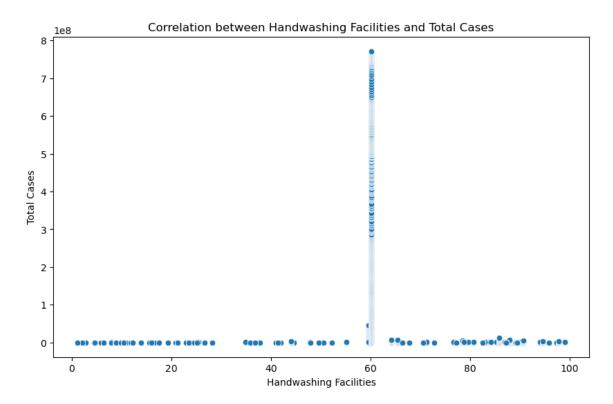
location	
Japan	18.493
Italy	16.240
Germany	15.957
Portugal	14.924
Greece	14.524
Latvia	14.136
Spain	13.799
Mean Age:	
location	
Japan	48.2
Italy	47.9
Germany	46.6
Portugal	46.2
Martinique	45.7
Spain	45.5
Greece	45.3



As is characteristic of the countries that have the highest number of 65-year-olds, they also have a higher average age

HANDWASHING

The Pearson correlation coefficient between "handwashing_facilities" and "total_cases" is ap proximately 0.0331. The p-value is very close to zero (9.81e-34), indicating that the correlation is statistically significant. However, the correlation coefficient is relatively low, suggesting a weak positive correlation between the availability of handwashing facilities and the total number of COVID-19 cases. This means that as handwashing facilities increase, there is a slight tendency for total cases to increase, but the relationship is not very strong.



Explored relationships between variables, e.g., 'Life Expectancy' and 'GDP per Capita.' the correlation coefficient between 'reproduction_rate' and 'new_cases' is approximately 0.0197. This indicates a very weak positive correlation between these two variables. In other words, there is a slight tendency for an increase in the reproduction rate to be associated with a slight increase in the number of new cases, but the relationship is not strong.

Analyzed the correlation between 'Age' and 'Total Cases,' including statistical analysis.

Correlation Coefficient: 0.047103495045963714

P-Value: 1.4928627922317368e-134

The correlation is statistically significant.

Mean Age: 30.344619107843815

Mean Total Cases: 3709936.6972507825 Standard Deviation Age: 9.070943013411519

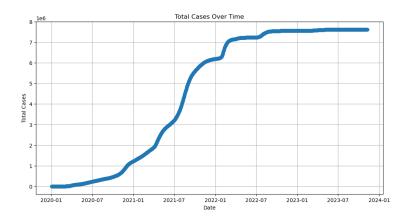
Standard Deviation Total Cases: 34038178.97808736

The correlation coefficient between 'median_age' and 'total_cases' is approximately 0.0471, and the p-value is very close to zero (1.49e-134). With such a low p-value, we can conclude that the correlation is statistically significant.

Correlation Coefficient (r): The positive correlation coefficient indicates a weak positive relationship between the median age of a population and the total number of COVID-19 cases. However, the strength of the correlation is relatively low. Statistical Significance: The very low p-value (close to zero) suggests that the observed correlation is unlikely to be due to random chance. Therefore, we have evidence to reject the null hypothesis and accept the alternative hypothesis that there is a significant correlation.

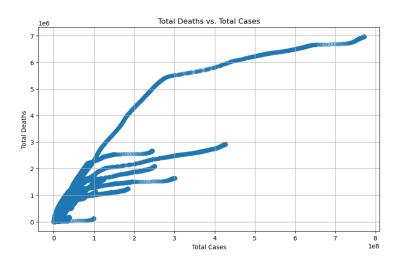
Total Cases Over Time:

Line plot showing the total cases over time.



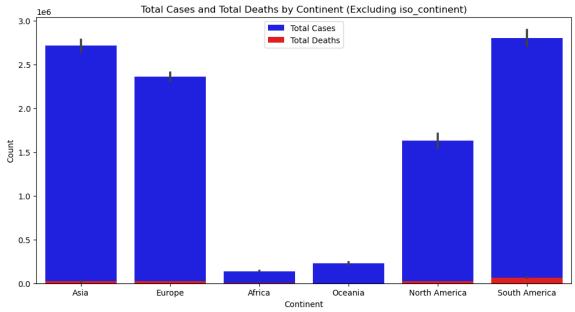
Total Deaths vs. Total Cases:

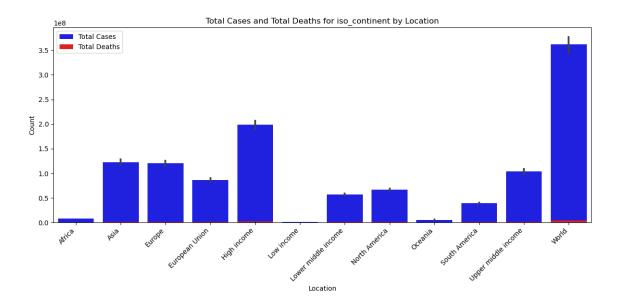
Scatter plot showing the relationship between total deaths and total cases.



Total Cases and Total Deaths by Continent:

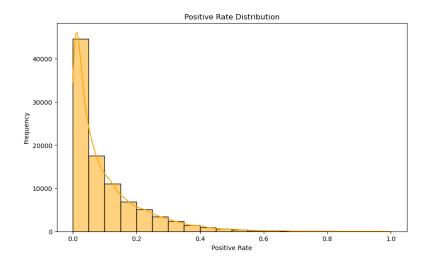
Bar plot showing the total cases and total deaths for each continent.





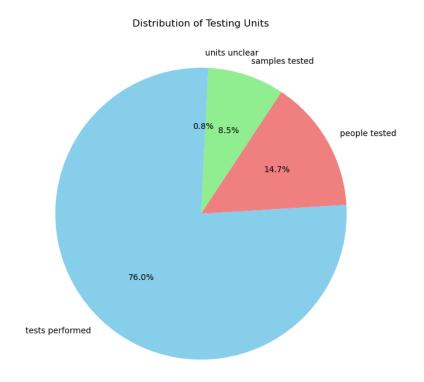
Positive Rate Distribution:

Histogram showing the distribution of positive rates.



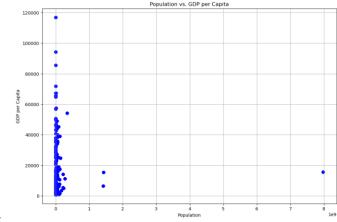
Testing Units Distribution:

Pie chart showing the distribution of testing units.

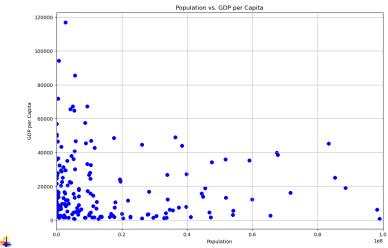


Population vs. GDP per Capita:

Scatter plot showing the relationship between population and GDP per capita.



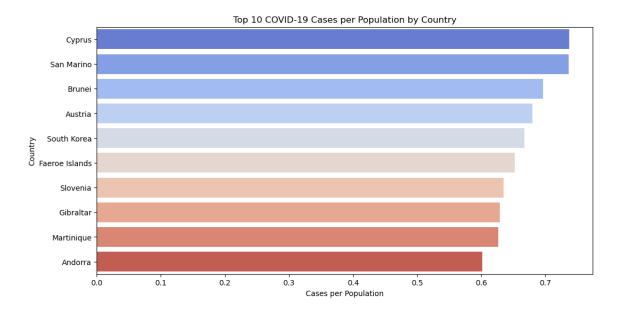
After plotting, I realized that the criteria was not correct and drew a new chart again with the new criteria.



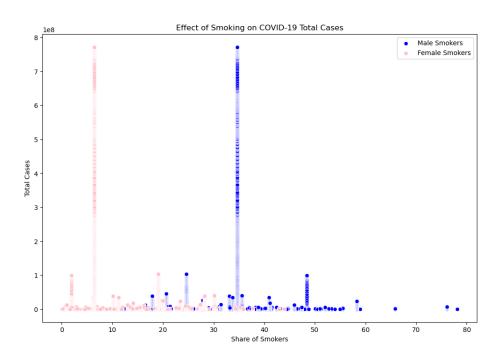
Well, as the correlation function showed, there is no linear relationship

Top COVID-19 cases:

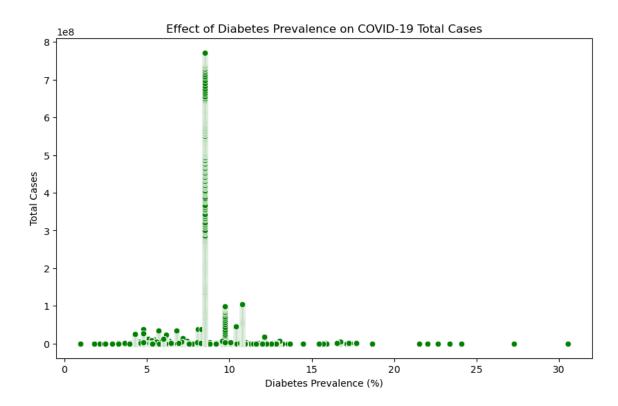
creating a bar chart to visualize the top 10 countries with the highest COVID-19 cases per population ratio



SMOKING



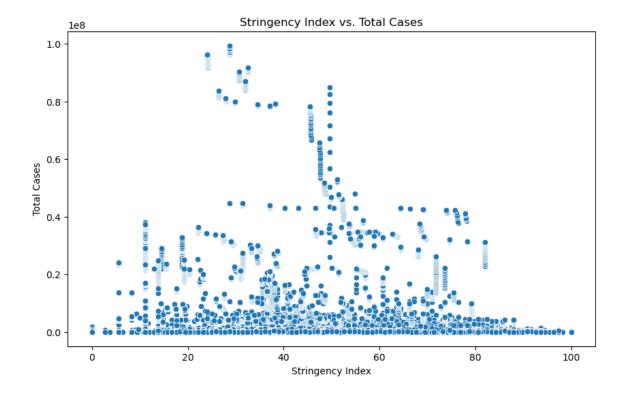
DIABETES



Correlation between Diabetes Prevalence and Total COVID-19 Cases: -0.0057131158 68240373

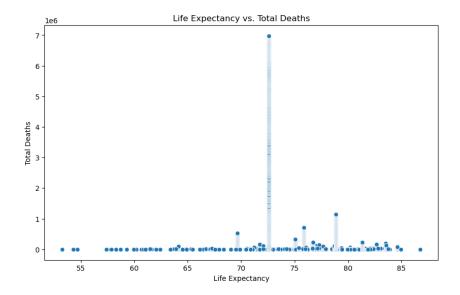
Correlation between Male Smokers Percentage and Total COVID-19 Cases: 0.0036120134780312355

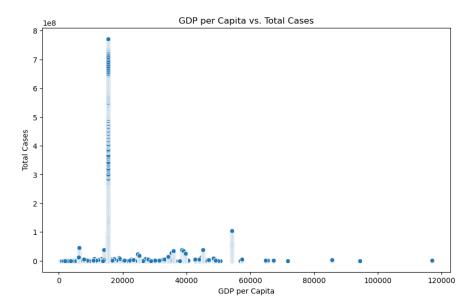
cardiovasc death rate	0.425771
hospital beds per thousand	0.347238
handwashing facilities	0.327041
excess mortality cumulative per million	0.248458
diabetes prevalence	0.201121
extreme poverty	0.192306



According to the plot, it can be understood that there is no relationship:

 The correlation between 'stringency_index' and 'total_cases' is -0.0631996766158127.

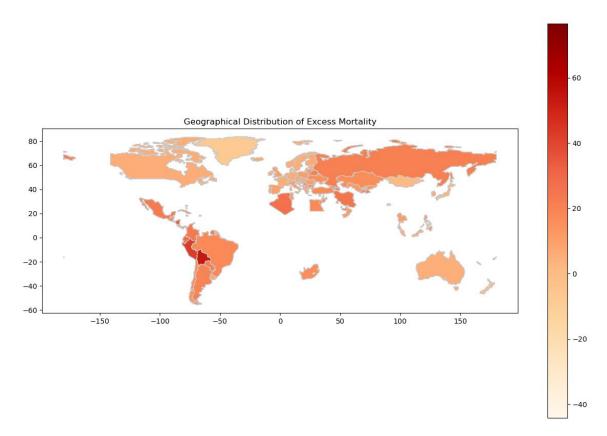




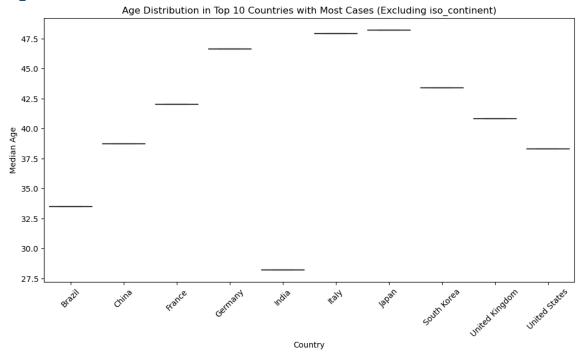
- The mean of 'life_expectancy' is approximately 73.72, and the mean of 'gdp_per_capita' is approximately 17418.15.
- It seems that there is no relationship and the plot is shown only on very high values which is the Mean.

Excess mortality

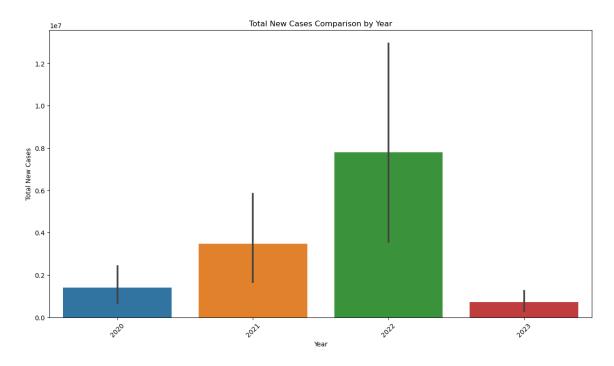
- the excess mortality can be calculated by taking the observed number of deaths during
 a specific period and subtracting the expected number of deaths based on historical
 data for the same period. This measure provides a more comprehensive
 understanding of the impact of a health crisis on mortality, especially when there may
 be underreporting or misclassification of causes of death.
- In the context of COVID-19, excess mortality has been used to assess the true impact of the virus, considering both confirmed COVID-19 deaths and any additional deaths that may be related to the pandemic's indirect effects.
- In summary, a higher excess mortality suggests that the observed number of deaths
 exceeds what would be expected under normal circumstances, indicating the broader
 impact of a health crisis on mortality.



Age



BY Year



3. Summary

This report delves into a detailed analysis of a COVID-19 dataset, encompassing 355,434 entries and 66 columns. The exploration begins with an initial inspection, revealing missing values and anomalies, leading to meticulous data cleaning strategies.

Initial Data Inspection:

- The dataset spans until 2023-11-09, with the last day data reported for North Korea showing NaN, indicating potential data unavailability.
- Investigation identifies 'North Korea' and 'Turkmenistan' as countries with NaN total cases on the last date; these entries are subsequently dropped.

Missing Values Analysis:

- Various columns exhibit missing values, with 'reproduction_rate,' 'icu_patients,' and 'stringency_index' having notably high percentages.
- Imputation and handling strategies are applied based on the nature of the data, considering the meaningfulness of null values.

Handling Missing Data:

- The 'continent' column is updated using the 'iso_continent' prefix for specific countries.
- Missing values in 'total_cases' are replaced with zeros, and countries with '0' total cases on their last reported day are identified and handled accordingly.

Correlation Analysis:

- Correlation coefficients are calculated, revealing strong positive correlations between 'total_cases' and 'total_deaths,' 'total_cases' and 'total_vaccinations,' among others.
- Features with high correlation to 'total_cases' are identified, including 'total_deaths,' 'total_boosters,' and 'total_tests.'

Exploration of Specific Variables:

- In-depth analyses are conducted on variables such as 'reproduction_rate,' 'icu_patients,' 'hosp_patients,' and 'weekly_icu_admissions.'
- Hospitalization spikes during March 2022 and January 2021 prompt further investigation into potential contributing factors.

Data Visualization:

- A wide array of visualizations, including line charts, bar plots, scatter plots, and histograms, offer insights into the distribution and relationships of key variables.
- Geographical analysis using maps provides a global perspective on COVID-19 statistics.

Feature Removal:

• Features with high percentages of missing values or limited relevance are dropped post-analysis.

Health and Demographic Analysis:

• Life expectancy, age distribution, and other demographic factors are explored, revealing intriguing insights.

Excess Mortality:

• Excess mortality is discussed as a crucial metric for assessing the broader impact of the health crisis.

Conclusion:

 The report concludes with a summary of the dataset's characteristics and a thorough exploration of various dimensions, providing a comprehensive understanding of the COVID-19 data.

This report not only offers a meticulous exploration of the dataset but also demonstrates a thoughtful approach to handling missing data and deriving meaningful insights from diverse variables. The combination of statistical analyses and visualizations enhances the clarity of findings, making this report a valuable resource for understanding the dynamics of the COVID-19 dataset.