IST 652 : Scripting for Data Analysis

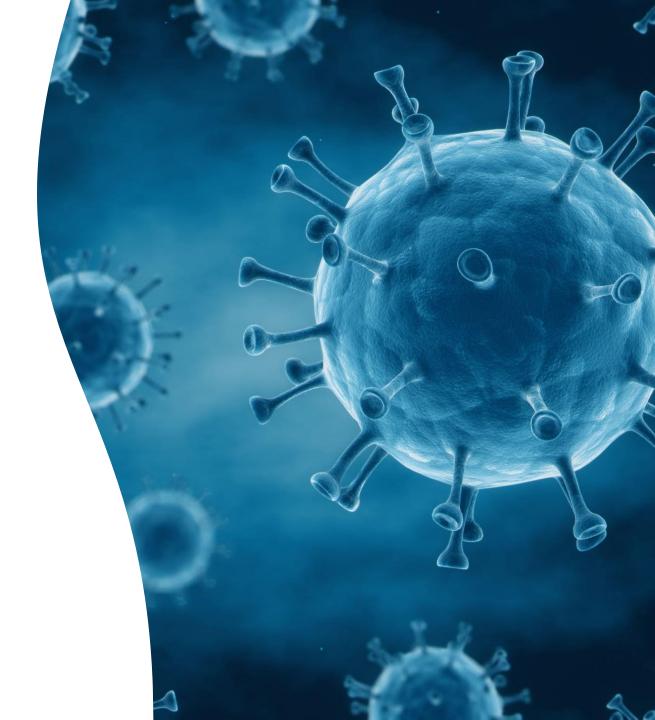
COVID-19 Deaths & Mentions in the United States

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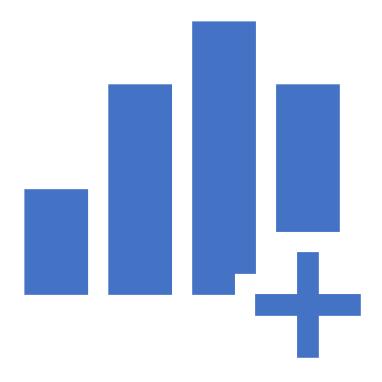
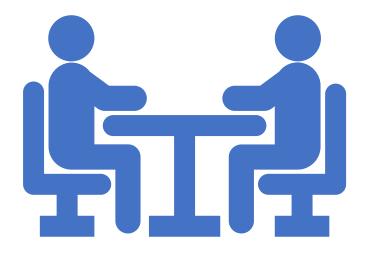


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Introduction

- When we sat down to discuss the dataset for our final project, we all agreed to do something that was health related and something that was recent
- The pandemic left many disturbances today including the underlying health conditions affiliated with COVID. This in return opens a wide range of possibilities in order to understand future precautionary measures
- In this presentation we will cover how COVID-19 deaths impacting a range of factors in the United States

Objectives

Visualization for Communication:

• Use visualizations (such as histograms, line plots, and heatmaps) to effectively communicate trends and patterns to a diverse audience, making the analysis accessible and informative.

Predictive Modeling

 Implement predictive models into our analysis and forecast future trends in COVID-19 deaths based on historical data.

Comparative Analysis:

• Compare COVID-19 death rates across different health conditions, states, or demographics to draw insights into the effectiveness of public health measures and healthcare systems.

Understanding the Impact:

• Analyze the overall impact of COVID-19 on mortality rates, identifying trends and patterns in the number of deaths over the pandemic.

Geospatial Analysis:

 Explore geographic variations in COVID-19 deaths, examining how different states have been affected using heatmaps

Demographic Patterns:

• Investigate demographic factors such as age and underlying health conditions to understand how different populations are affected by COVID-19

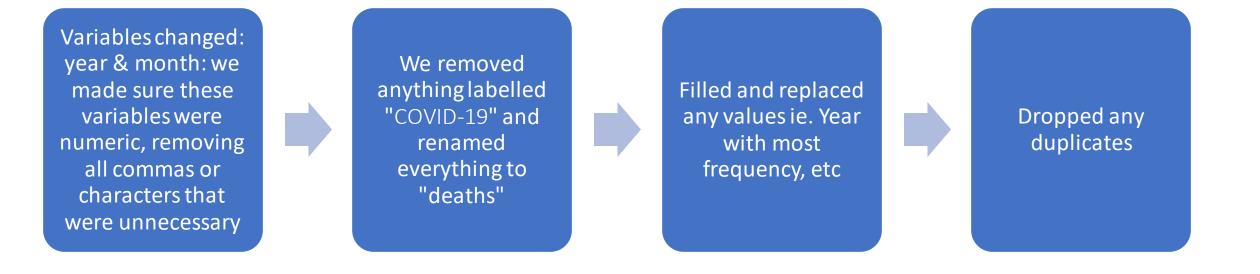
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Data columns (total 22 columns):

Data	Cotumns (total 22 C	o cullins / .	
#	Column	Non-Null Count	Dtype
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1	id	621000 non-null	object
2	position	621000 non-null	int64
3	created_at	621000 non-null	int64
4	created_meta	0 non-null	object
5	updated_at	621000 non-null	int64
6	updated_meta	0 non-null	object
7	meta	621000 non-null	-
8	Data As Of	621000 non-null	object
9	Start Date	621000 non-null	object
10	End Date	621000 non-null	object
11	Group	621000 non-null	object
12	Year	608580 non-null	object
13	Month	558900 non-null	object
14	State	621000 non-null	object
15	Condition Group	621000 non-null	object
16	Condition	621000 non-null	object
17	ICD10_codes	621000 non-null	object
18	Age Group	621000 non-null	object
19	COVID-19 Deaths	437551 non-null	object
20	Number of Mentions		object
21	Flag	183449 non-null	object

Data Description

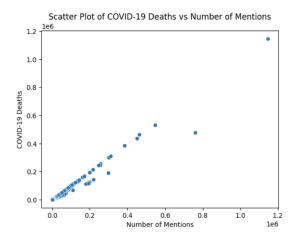
- Source: The dataset was sourced from Data.gov
- The dataset has 124,200 rows and 22 columns
- **Timeframe**: Covers data from [start date] to [end date], providing a comprehensive view of the COVID-19 impact over time.
- Scope and Scale: Comprises over 600,000 records, reflecting a wide range of demographic and clinical data points across the United States.
- Key Variables:
- **State**: Geographical location within the United States, excluding entries labeled as 'United States' to focus on individual states.
- Condition Group and Condition: Categorization of COVID-19 associated health conditions as per ICD10 clinical codes.
- **COVID-19 Deaths:** Recorded fatalities attributed to COVID-19, requiring conversion from object to numeric data type for analysis.
- **Number of Mentions:** Frequency count of specific conditions or keywords in the dataset, potentially linked to reported deaths.
- Age Group: Demographic segmentation of data, which can provide insights into the age-related impact of the pandemic.

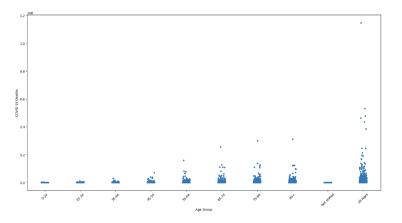
Data Cleaning



Data Exploration: Scatter Plot & Stripp Plot

- Explored the data with df.head(), df.info(), df.describe()
- Created a scatter plot for Deaths Vs Number of cases to understand the relationship between these variables
- Plotted stripp plot to understand how age group factor has an impact on Covid 19 deaths





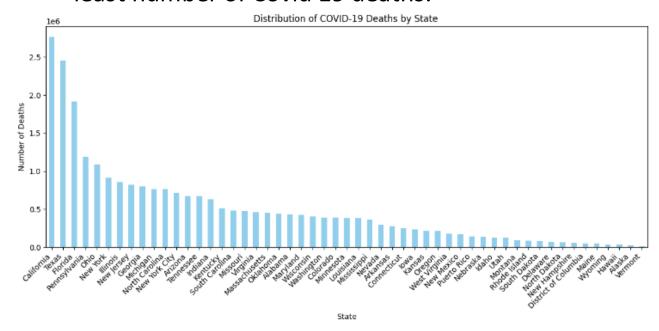
Age Filtering

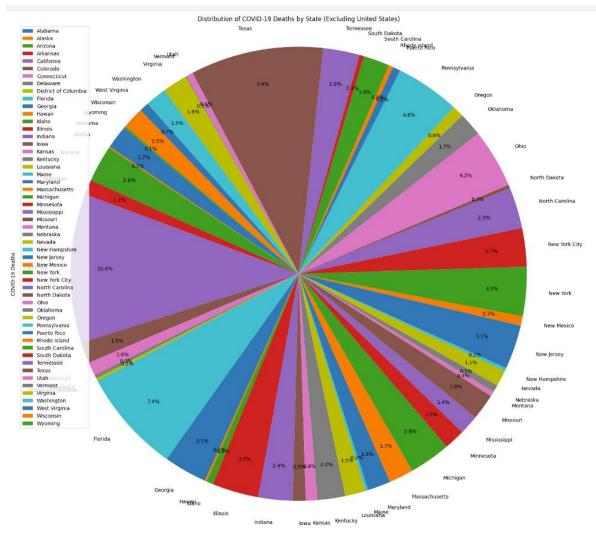
- Performed filtering on the Age group variable
- Age filtering for above age of 65

_meta	updated_at	updated_meta	meta	Data As Of	Start Date	 Year	Month	State	Condition Group	Condition	ICD10_codes	Age Group	COVID- 19 Deaths	Number of Mentions	Flag
None	1695825684	None	{}	2023-09- 24T00:00:00	2020-01- 01T00:00:00	 None	None	United States	Respiratory diseases	Influenza and pneumonia	J09-J18	65-74	129005.0	133088	None
None	1695825684	None	{}	2023-09- 24T00:00:00	2020-01- 01T00:00:00	 None	None	United States	Respiratory diseases	Influenza and pneumonia	J09-J18	85+	121119.0	123018	None
None	1695825684	None	{}	2023-09- 24T00:00:00	2020-01- 01T00:00:00	 None	None	United States	Respiratory diseases	Chronic lower respiratory diseases	J40-J47	65-74	27920.0	29359	None
None	1695825684	None	{}	2023-09- 24T00:00:00	2020-01- 01T00:00:00	 None	None	United States	Respiratory diseases	Chronic lower respiratory diseases	J40-J47	85+	27866.0	28796	None
None	1695825684	None	{}	2023-09- 24T00:00:00	2020-01- 01T00:00:00	 None	None	United States	Respiratory diseases	Adult respiratory distress syndrome	J80	65-74	30138.0	30138	None

Pie Chart & Bar Graph

- Pie chart and bar graph shows the distribution of Deaths by states.
- California, Texas, followed by Florida recorded most number of deaths. Where as Alaska, Vermont have least number of Covid 19 deaths.





Grouping Data

Counted Condition Group

Condition Group	
All other conditions and causes (residual)	27000
Alzheimer disease	27000
COVID-19	27000
Circulatory diseases	189000
Diabetes	27000
Intentional and unintentional injury, poisoning, and other adverse events	27000
Malignant neoplasms	27000
Obesity	27000
Renal failure	27000
Respiratory diseases	162000
Sepsis	27000
Vascular and unspecified dementia	27000
dtype: int64	

Number of Mentions Per Age Group

Age Group	
0-24	1.556827
25-34	5.398871
35-44	15.637929
45-54	43.152662
55-64	107.697316
65-74	178.955195
75-84	203.862396
85+	196.574813
All Ages	622.197288
Not stated	0.005286
Name: Number	of Mentions, dtype: float64

Number of Deaths by State

[14]: State

[1+]	State	
	Alabama	432004.0
	Alaska	25563.0
	Arizona	675395.0
	Arkansas	277702.0
	California	2765450.0
	Colorado	388689.0
	Connecticut	248369.0
	Delaware	69242.0
	District of Columbia	52580.0
	Florida	1915568.0
	Georgia	797183.0
	Hawaii	35550.0
	Idaho	127624.0
	Illinois	857240.0
	Indiana	630686.0
	Iowa	238231.0
	Kansas	214046.0
	Kentucky	512253.0
	Louisiana	381519.0
	Maine	49061.0
	Maryland	423271.0
	Massachusetts	454212.0
	Michigan	766221.0
	Minnesota	383921.0
	Mississippi	365180.0
	Missouri	477690.0
	Montana	92001.0
	Nebraska	134276.0
	Nevada	297838.0
	New Hampshire	59426.0
	New Jersey	818458.0
	New Mexico	174399.0
	New York	912528.0
	New York City	713151.0
	North Carolina	761941.0
	North Dakota	65317.0
	Ohio	1090446.0
	Oklahoma	438130.0
	Oregon	213144.0
	Pennsylvania	1190005.0
	Puerto Rico	142259.0
	Rhode Island	84425.0
	South Carolina	483097.0
	South Dakota	81556.0
	Tennessee	673816.0
	Texas	2453758.0
	United States	26501616.0
	Utah	126856.0
	Vermont	18419.0
	Virginia	460649.0
	Washington	390447.0
	West Virginia	175517.0
	Wisconsin	402985.0
	Wyoming	36806.0
	Name: COVID-19 Deaths,	
	name. Covid-15 Deaths,	acype, 110aco+

- Grouping data by 'State' and calculating total COVID-19 deaths per state
- Grouping data by 'Condition Group' and counting the occurrences of each group
- Grouping data by 'Age Group' and calculating average number of mentions per age group

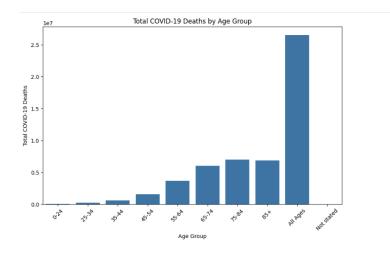
Analysis

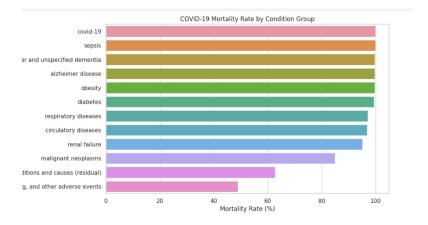
Bar Chart: Total COVID-19 Deaths by Age Group

- This bar chart illustrates the distribution of COVID-19 related deaths across different age groups.
- The data suggests that the impact of COVID-19 on mortality rates increases with age, with the highest number of deaths occurring in the oldest age bracket.

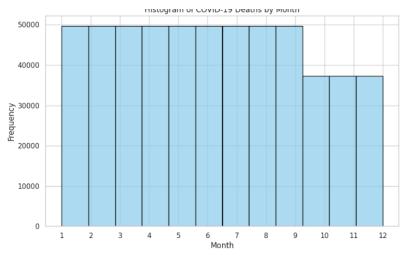
Stacked Bar Chart: COVID-19 Mortality Rate by Condition Group

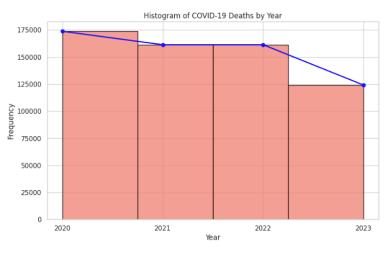
- The stacked bar chart presents the mortality rate of patients with COVID-19 in conjunction with various underlying health conditions or comorbidities.
- The length of each bar represents the percentage of the mortality rate attributed to each condition group, providing insight into which health issues are most commonly associated with fatal COVID-19 outcomes.
- This visualization emphasizes the significant risk factors, such as cardiovascular diseases and diabetes, contributing to COVID-19 mortality.





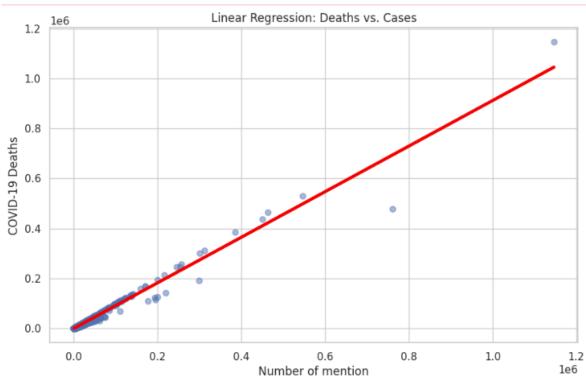
Histogram Analysis





- For the "Histogram of COVID-19 Deaths by Month," you can observe the distribution of COVID-19 deaths across different months. The x-axis represents the months, and the y-axis represents the frequency of deaths.
- For the "Histogram of COVID-19 Deaths by Year," the histogram gives an overview of the distribution of deaths across different years, while the line plot shows the trend or pattern of COVID-19 deaths over the years. The line plot connects the high points of the histogram, providing a visual representation of the variation in the number of deaths each year.

Linear Regression Analysis

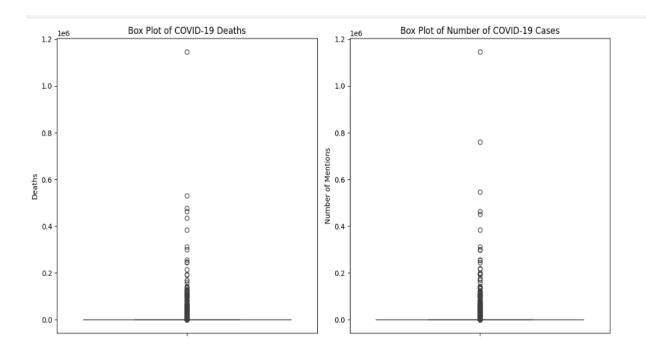


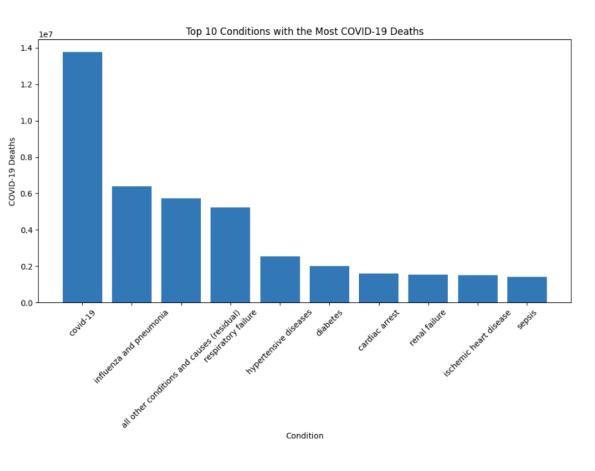
Intercept: 0.6028922878045648 Coefficient: 0.9117763436901557

- Predictive model that showcases how deaths and mentions are closely related to the data's average
- As deaths increase, the number of mention increases (and vice versa), creating a positive linear regression line

Outlier Analysis-Box Plot

- Both COVID-19 deaths and cases have a skewed distribution with most data points concentrated at the lower end.
- The median values are closer to the bottom of the data range, indicating a lower central tendency for both deaths and cases.
- There are numerous outliers for both deaths and cases, signifying instances of very high numbers that deviate from the typical values.
- The variance in the data is substantial, with the bulk of observations being low, but with some significantly high values.



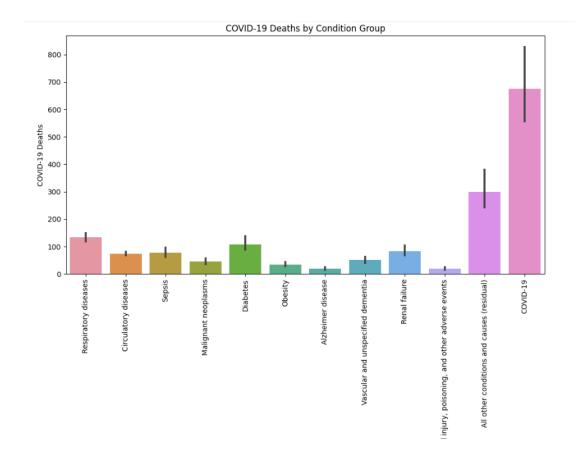


Hypothesis Testing & Frequency Distribution

- Comparing the COVID-19 deaths to condition variable in the dataset
- The ANOVA results show a high F-statistic of approximately 53.98 and a very low p-value (around 3.49×10–2373.49×10–237). This indicates a statistically significant difference in COVID-19 deaths across different condition groups in our dataset, allowing to reject the null hypothesis that there is no difference between group means.
- In Frequency Distribution, the bar chart displays the top 10 conditions associated with COVID-19 deaths, with "Pneumonia" accounting for the highest number of deaths, followed by and "Influenza".
- The decreasing order indicating the relative impact of each condition on COVID-19 mortality.

ANOVA F-statistic: 53.980576930153816

p-value: 3.498008065053823e-237



Plot of Deaths By Condition Group

- **Chart Analysis**: Demonstrates COVID-19 fatalities by comorbid conditions.
- Insights:
- 'COVID-19' is the predominant cause of recorded deaths.
- 'Alzheimer's' and 'dementia' are notable for their high impact.
- Significant fatalities also occurred in patients with 'Respiratory Diseases', 'Diabetes', and 'Heart Conditions'.

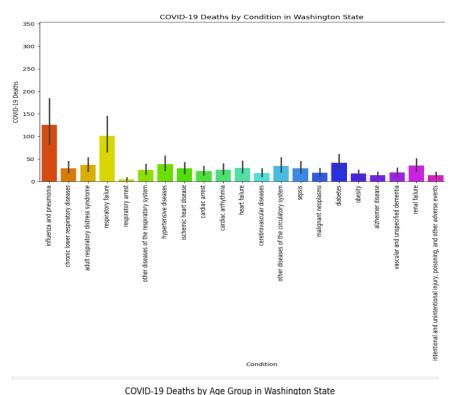
Analysis of Washington State

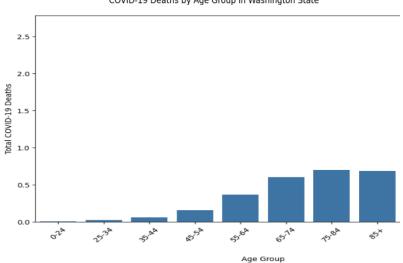
First Plot: COVID-19 Deaths by Condition in Washington State

- This bar chart provides an in-depth look at the comorbidities or underlying conditions associated with COVID-19 fatalities in Washington State.
- Each bar represents a different medical condition and is color-coded for clear differentiation.
- The plot highlights which conditions have been most frequently associated with death cases, with the tallest bars indicating the highest number of deaths.
- Notably, conditions such as 'pneumonia' and 'sepsis' show significantly higher mortality figures, suggesting their severe impact on COVID-19 patients.

Second Plot: COVID-19 Deaths by Age Group in Washington State

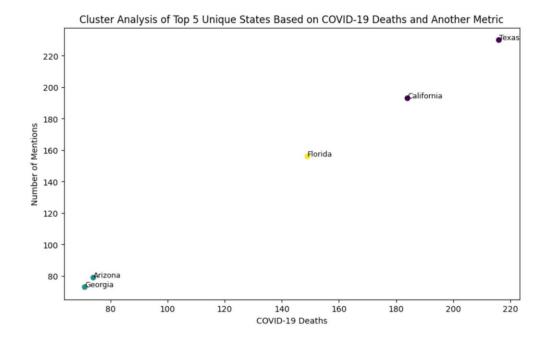
- This histogram breaks down the total number of COVID-19 deaths by age group within the state, offering a clear demographic perspective.
- A stark increase in deaths is observed among the older populations, with the highest number occurring in the '65+' age category.
- The distribution signifies the heightened vulnerability of the elderly to COVID-19, underscoring the need for targeted protective measures for senior citizens.

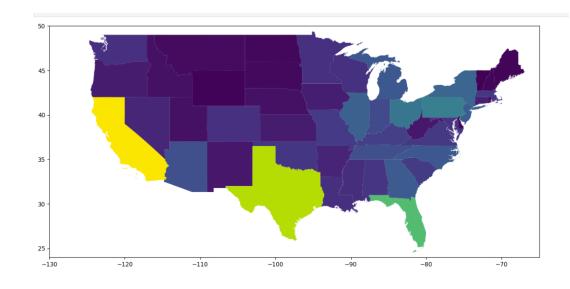


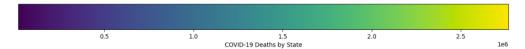


Cluster Analysis

- **Visual Overview**: The scatter plot displays a comparison of the top five states by COVID-19 deaths against the number of times COVID-19 is mentioned in official records.
- Axes Interpretation:
- **X-Axis**: Represents the count of COVID-19 deaths.
- **Y-Axis**: Denotes the number of mentions of COVID-19 in documents.
- Insights:
- Texas & California: High impact, with large numbers of deaths and mentions.
- Florida: Notable for a high number of mentions relative to deaths.
- Georgia & Arizona: Lower impact, with fewer deaths and mentions.







Heatmap

- **Purpose of the Heatmap**: This map visualizes the intensity of COVID-19 deaths across the United States, using color coding to represent varying levels of impact.
- Color Interpretation:
- Dark Purple: Lower numbers of reported deaths.
- Yellow-Green: Higher numbers of reported deaths.
- The scale at the bottom translates the color gradient into the actual number of deaths.
- Insight:
- States with darker shades show relatively fewer deaths, while those in yellow-green have reported higher death counts.
- The color gradient provides an at-a-glance understanding of geographical trends in COVID-19 mortality rates.
- **Usage**: Such a heatmap can quickly convey regional differences and identify areas with the most significant health impact from the pandemic, which can be crucial for targeted public health responses.

Challenges



Large Dataset: With over 600,000 entries, processing and analyzing the dataset can be computationally intensive and may require optimization techniques or more robust hardware.



Data Type Discrepancies:
Columns like 'COVID-19
Deaths' and 'Number of
Mentions' being in an
object data type instead of
numeric, necessitating
type conversion and error
handling.



Missing Values: Significant numbers of missing entries in crucial columns, such as 'COVID-19 Deaths' and 'Number of Mentions', require careful handling to avoid bias.



Geospatial Data
Integration: Difficulties in
integrating geospatial data
due to format issues (e.g.,
shapefiles not being
directly accessible or
compatible).



Clarity of Visualizations:
Ensuring that complex
data is represented in a
way that is
understandable and
visually clear to the
audience.



Choice of Appropriate
Visuals: Determining the
most effective type of
visualization for the data
at hand, such as choosing
between heatmaps, bar
charts, or cluster
diagrams.

Conclusion

- Our analysis revealed distinct temporal trends, highlighting periods of heightened mortality rates. Additionally, we observed intriguing geographic variations, emphasizing the importance of localized interventions and healthcare strategies.
- By showcasing the demographic factors, we identified specific populations that faced elevated risks. Understanding the impact of age, gender, and underlying health conditions is crucial for targeted public health efforts.
- Our use of Python's visualization capabilities not only facilitated a deeper understanding of the data but also made complex information accessible to a broader audience. Effective communication through visuals is crucial for informed decision-making.



Questions?

THANK YOU!