



University of Colorado  
Boulder

# COVID-19 Comprehensive Study

## CSPB-4502 Data Mining Project

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# Introduction

Our project focuses on analyzing data surrounding COVID-19, including deaths, vaccination rates, and cases viewed on a per-county basis, covering dates from 1/1/2020 to 9/17/2022. Our group decided to use the COVID-19 Open Data by Google<sup>1</sup>. The intriguing questions we aim to answer are:

- How do local variables, such as temperature and mobility rates, affect COVID rates?
- Did high vaccination rates help mitigate the deaths and spread of the virus?
- Understand the impact of detailed weather variables on COVID-19 rates
- How do regional outbreaks occur and do they spread between states/counties?

<sup>1</sup><https://health.google.com/covid-19/open-data>



# Tools and Methodology

- Overleaf
- Git/GitHub
- Python:
  - Pandas
  - NumPy
  - Geopandas
  - Sklearn
  - Matplotlib
- Agile Method:
  - Weekly check ins
  - Dedicated time to work individually and as a group
  - Shared goals for the week
  - Helped others troubleshoot obstacles



# Data Preprocessing

## Data Transformation:

- Standardized incidence rate per 100,000
- Min-Max Scaling implemented

## Data Repositories:

- Primary GitHub repository
- Storage of all county CSVs for analysis and a subset for testing data mining strategies.

## Data Reduction:

- Focus on the U.S.
- Dimension reduction

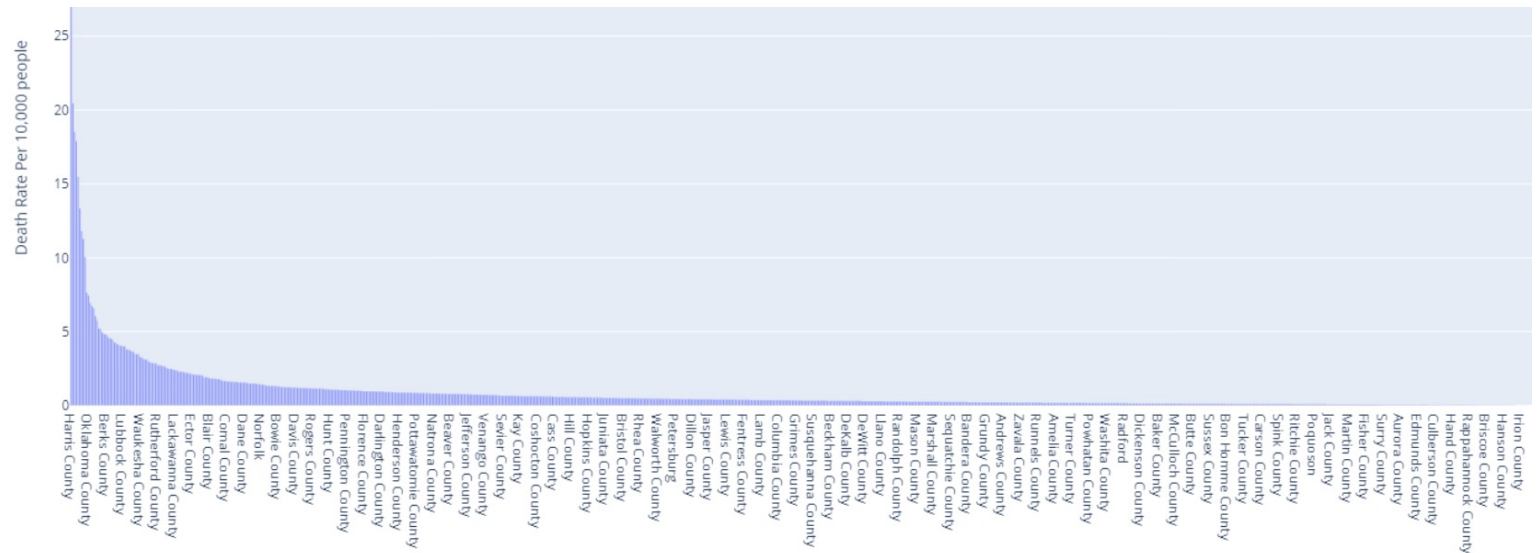
## Data Cleaning

- Made data consistent amongst all files
- Excluded attributes lacking too much data

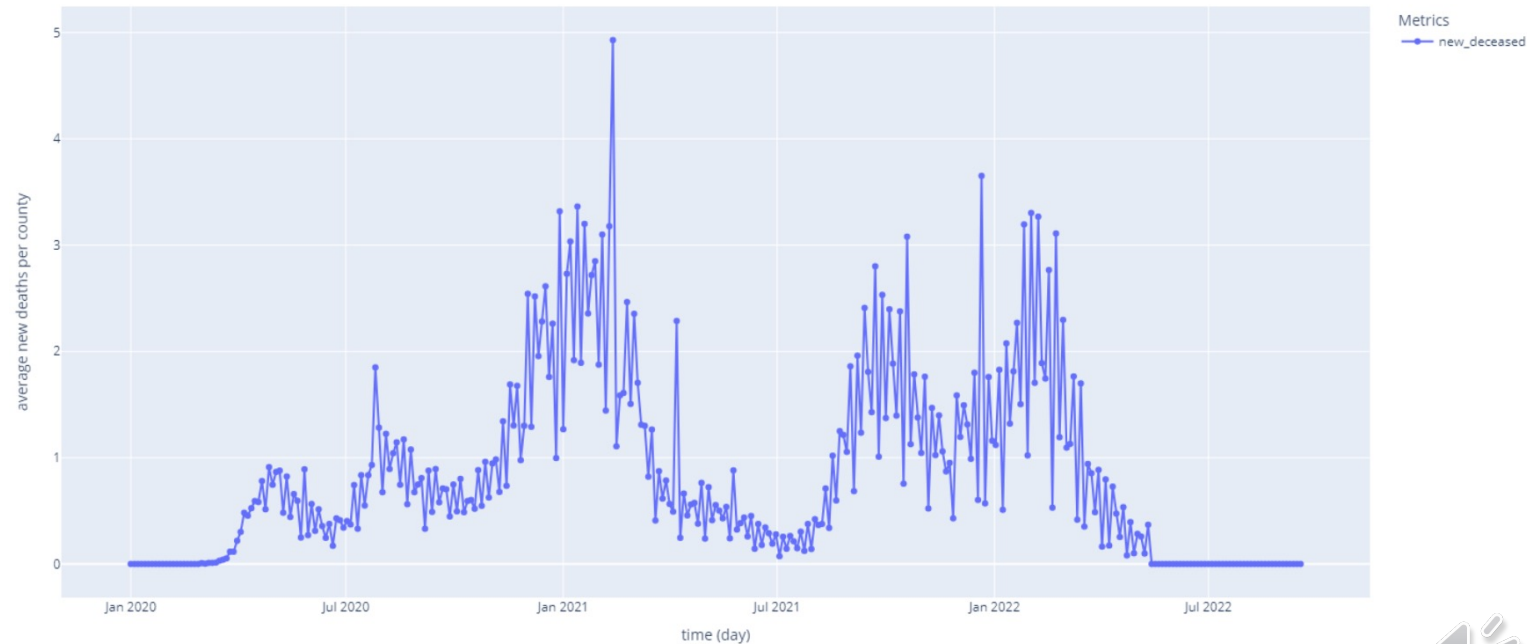


# Preliminary plots (data familiarization)

- Temporal graphs
- Bar charts



3-Day County Moving Average of Death Rate Over Time"



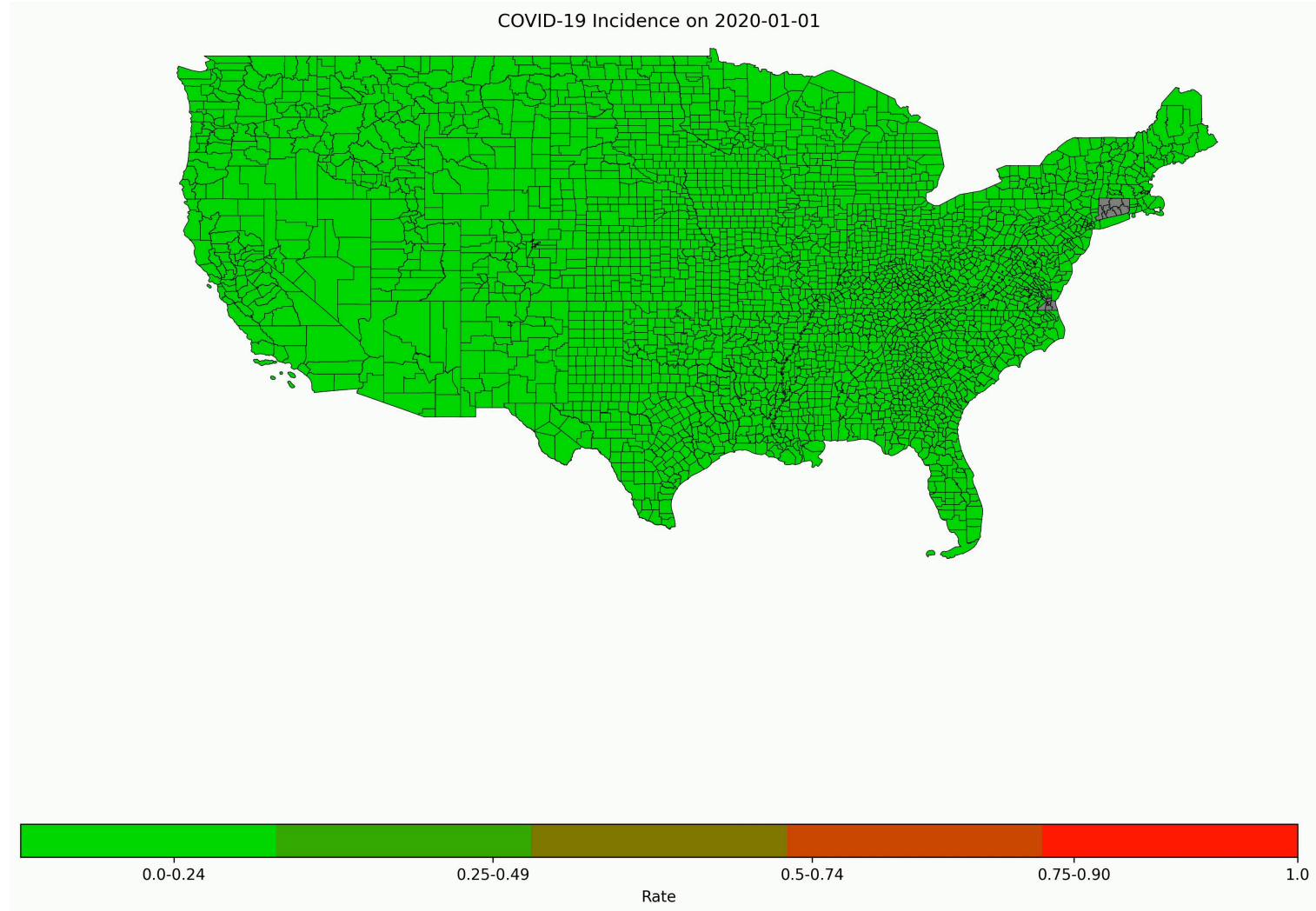
# Geospatial Analysis

- Used TIGER/Line Shapefiles to color each county to get more accurate representation of severity at a particular time interval
- Aggregated different timeframes ranging from 3-day intervals to 1-year intervals
- Binned to examine different scopes.
- We used the Python module Geopandas to facilitate plotting.



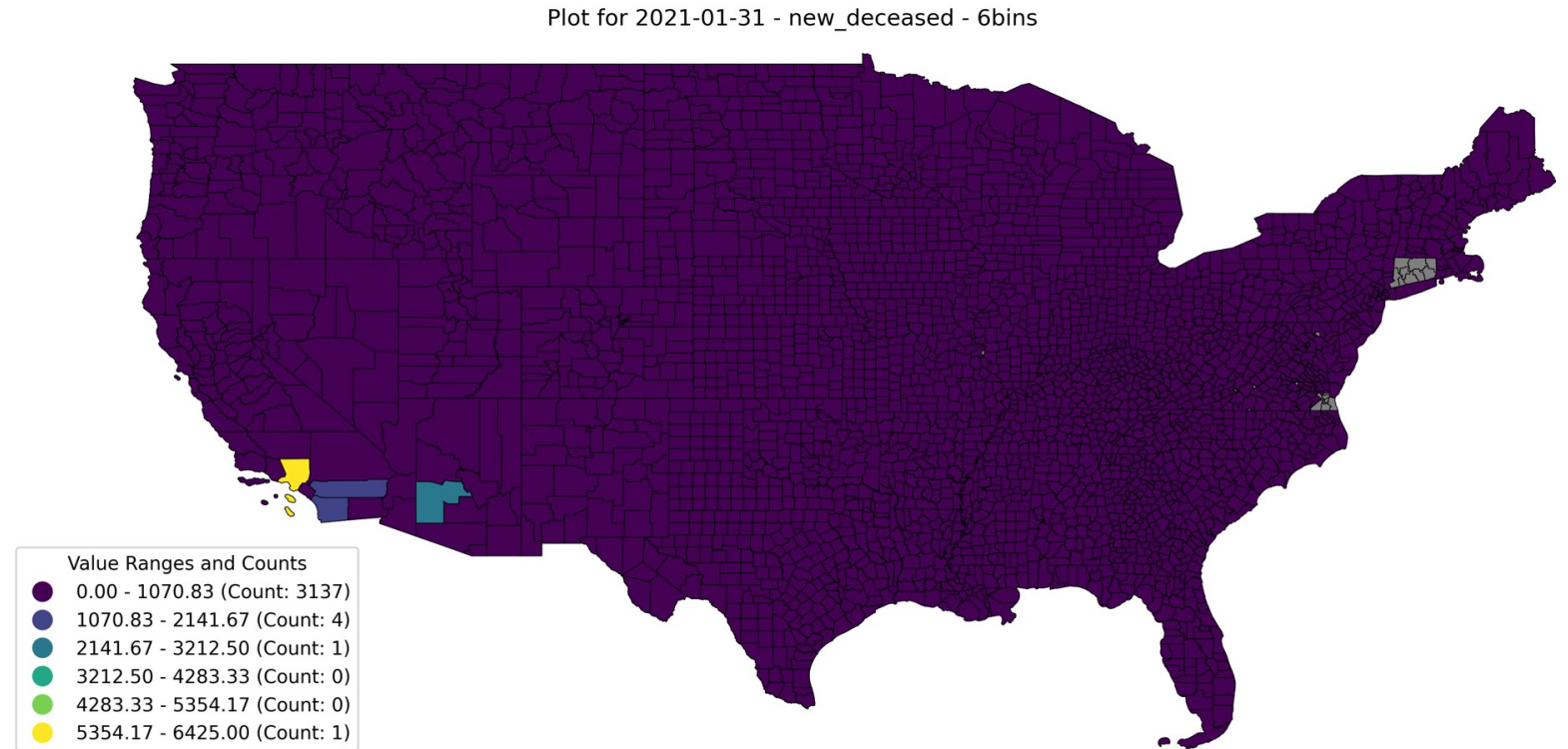
# Geospatial Progressive Timelapse

- Seem to have two waves of covid.
- Regions with outbreak seem to expand to others.
- Aggregation based on 17 days, per normalized per 100,000



# Geospatial on Worst County Outbreaks

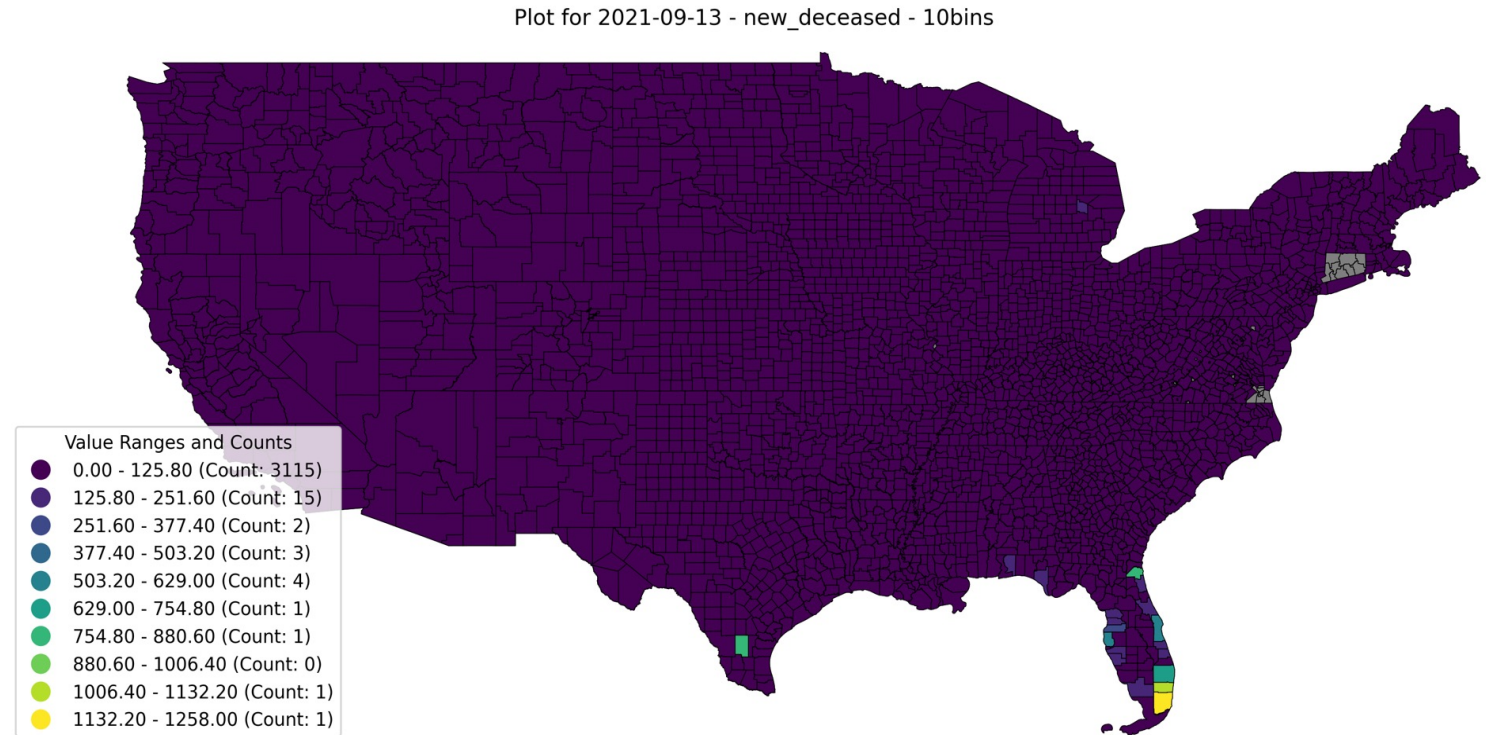
- During December 2020 through February 2021, the West Coast had a significant breakout.
- Used binning to emphasize contrast for worst counties





# Geospatial on Worst County Outbreaks (Cont.)

- Florida experienced a similar outbreak that was significantly more severe than other counties.
- Different aggregation timeframes led to identifying different outbreaks.



# Correlation Analysis

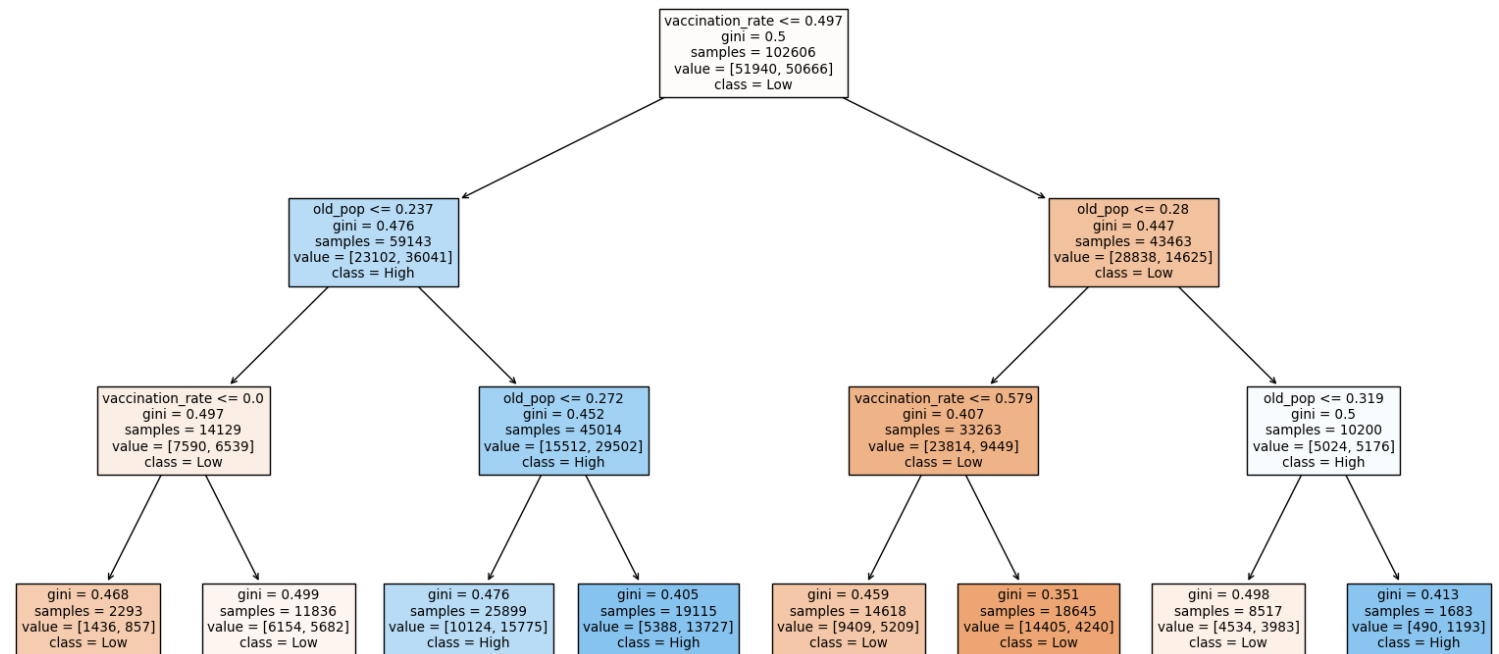
- Pearson Correlation Coefficients: Linear Relation.
- Moderate positive correlation between the new confirmed cases and new deceased cases.
- Mobility data show weak negative correlations with new confirmed cases
- Residential mobility shows a weak positive correlation with new confirmed cases.
- Temperature shows weak negative correlations with newly confirmed cases.

Variable	New Confirmed	New Deceased
new_confirmed	1.000000	0.240407
new_deceased	0.240407	1.000000
mobility_retail_and_recreation	-0.068893	-0.082005
mobility_grocery_and_pharmacy	-0.055497	-0.067243
mobility_parks	-0.070682	-0.062156
mobility_transit_stations	-0.099402	-0.102103
mobility_workplaces	-0.063318	-0.066253
mobility_residential	0.087180	0.099044
average_temperature_celsius	-0.029875	-0.014980
minimum_temperature_celsius	-0.025170	-0.009587
maximum_temperature_celsius	-0.034440	-0.021168
rainfall_mm	-0.005717	-0.003457
dew_point	-0.031608	-0.019824
relative_humidity	-0.008104	-0.011446



# Impact of Vaccination on COVID-19 Death Rates

- Decision Tree Classification
  - Feature variables: Vaccination Rate and Old Population Rate
  - Target variable: Death Rate
- Increased vaccination rate in counties with higher older population rates decreases the mortality rate in the county.
- This shows a correlation between vaccines and lower mortality rates for older populations.



# Comparative Regression Analysis with and without Lagged Variables - Ordinary Least Squares (OLS)

**Objective:** Examine the effectiveness of incorporating lagged variables into regression models for predicting COVID-19 cases.

**Methodology:** Two OLS regression models were developed; one with current environmental data and another enhanced with lagged environmental data from the previous week.

**Rationale:** To capture the delayed effects of environmental factors on COVID-19 transmission rates.



# Initial Model without Lagged Variables

- **Variables Used:** Average temperature, minimum temperature, maximum temperature, rainfall, and relative humidity.
- **Results Summary:**
  - R-squared: 0.082, indicating that about 8.2% of the variability in new confirmed COVID-19 cases per 1000 people is explained by the model.
  - Significant predictors: All initial environmental factors had a noticeable impact on COVID-19 case predictions.

OLS Regression Results						
=====						
Dep. Variable:	new_confirmed_per_1000	R-squared:	0.082			
Model:	OLS	Adj. R-squared:	0.082			
Method:	Least Squares	F-statistic:	8209.			
Date:	Mon, 29 Apr 2024	Prob (F-statistic):	0.00			
Time:	12:37:08	Log-Likelihood:	-1.1703e+06			
No. Observations:	457236	AIC:	2.341e+06			
Df Residuals:	457230	BIC:	2.341e+06			
Df Model:	5					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
const	2.2900	0.064	35.845	0.000	2.165	2.415
average_temperature_celsius	-0.8450	0.012	-71.910	0.000	-0.868	-0.822
minimum_temperature_celsius	0.3907	0.007	55.921	0.000	0.377	0.404
maximum_temperature_celsius	0.3750	0.006	67.120	0.000	0.364	0.386
rainfall_mm	-0.0105	0.000	-35.856	0.000	-0.011	-0.010
relative_humidity	0.0066	0.001	10.539	0.000	0.005	0.008
=====						
Omnibus:	597290.181	Durbin-Watson:	0.579			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	932470082.234			
Skew:	6.567	Prob(JB):	0.00			
Kurtosis:	223.844	Cond. No.	1.03e+03			
=====						



# Enhanced Model with Lagged Variables

- **New Variables:** Lagged confirmed cases and lagged average temperature were added.
- **Results Summary:**
  - R-squared Improved to 0.560, showing that 56% of the variability is now explained by the model, significantly enhancing predictive accuracy.
  - F-statistic: Increased to approximately 82,960, underscoring the model's robustness.

OLS Regression Results						
=====						
Dep. Variable:	new_confirmed_per_1000	R-squared:	0.560			
Model:	OLS	Adj. R-squared:	0.559			
Method:	Least Squares	F-statistic:	8.296e+04			
Date:	Mon, 29 Apr 2024	Prob (F-statistic):	0.00			
Time:	12:37:13	Log-Likelihood:	-1.0025e+06			
No. Observations:	457231	AIC:	2.005e+06			
Df Residuals:	457223	BIC:	2.005e+06			
Df Model:	7					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
const	0.4354	0.045	9.771	0.000	0.348	0.523
average_temperature_celsius	-0.2869	0.008	-35.068	0.000	-0.303	-0.271
minimum_temperature_celsius	0.1380	0.005	28.247	0.000	0.128	0.148
maximum_temperature_celsius	0.1299	0.004	33.393	0.000	0.122	0.138
rainfall_mm	-0.0039	0.000	-18.990	0.000	-0.004	-0.003
relative_humidity	0.0034	0.000	7.742	0.000	0.003	0.004
lagged_new_confirmed_per_1000	0.7256	0.001	703.698	0.000	0.724	0.728
lagged_avg_temp	0.0041	0.001	4.858	0.000	0.002	0.006
=====						
Omnibus:	579431.758	Durbin-Watson:	2.123			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	2994226686.607			
Skew:	5.797	Prob(JB):	0.00			
Kurtosis:	399.273	Cond. No.	1.06e+03			
=====						



# Comparative Analysis and Model Improvement

- **Improvements Noted:**

- Adjusted R-squared increased dramatically from 0.082 in the initial model to 0.560 in the enhanced model.
- The Durbin-Watson statistic improved, indicating reduced autocorrelation among residuals.

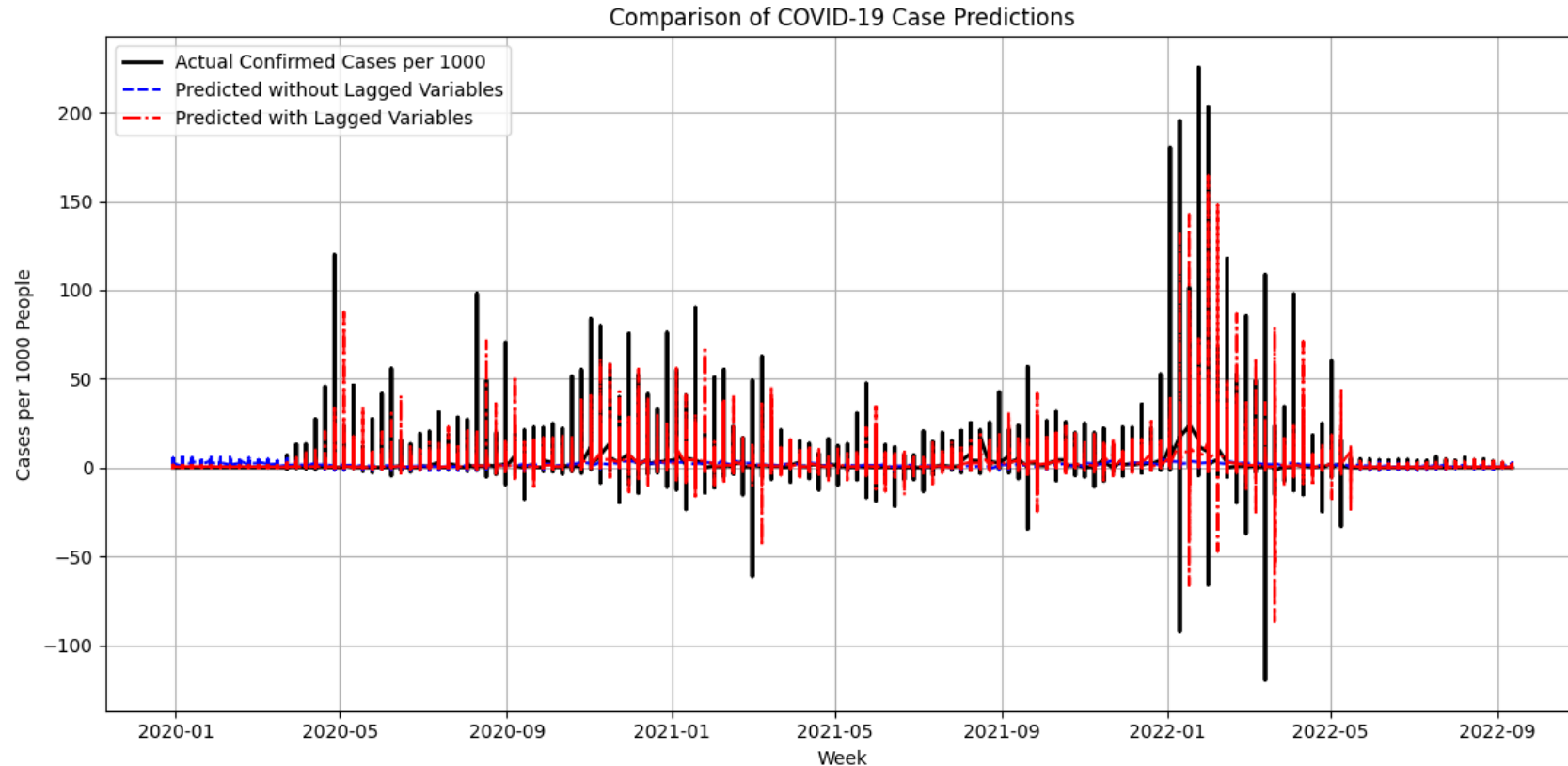
- **Diagnostics Issues:**

- Omnibus and Jarque-Bera tests still indicate non-normal distribution of residuals, suggesting the presence of outliers or model misspecification.
- Next Steps: Consideration of more sophisticated time-series models or transformations of the dependent variable to further refine model accuracy.





# Visual Comparison and Conclusion



- **Conclusion:** The inclusion of lagged variables significantly improves the model's ability to predict new COVID-19 cases, highlighting the importance of considering temporal dynamics in epidemiological modeling

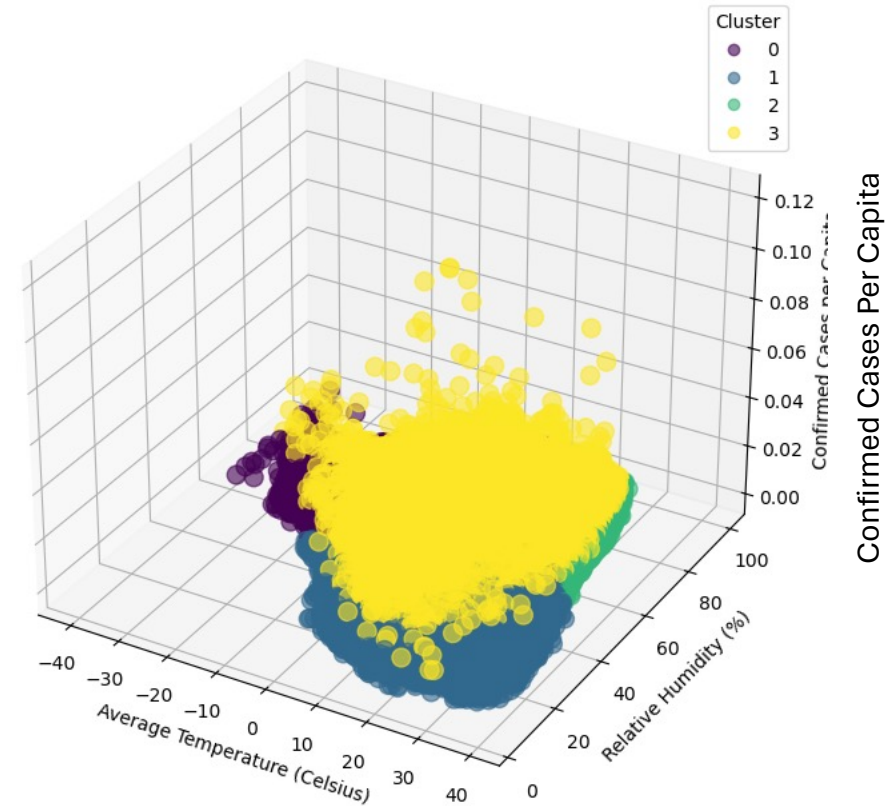




# Impact of Weather on COVID-19 Rates

- k-means clustering
  - 4 clusters based in inertia
- Decrease in the transmissibility of COVID-19 at temperatures exceeding 30°C, under 30°C, or under a 20% relative humidity.

3D Clusters by Temperature, Humidity, and COVID-19 Cases per Capita



# Knowledge Gained and Application

- Infections seem to expand from one region to others.
- There appeared to be at least two waves significant outbreaks.
- Higher vaccination rates are related to lower COVID-19 death rates.
- COVID-19 cases decrease in extreme climate.

