

## INTRODUCTION

This Covid 19 Data Analysis project aims to determine the reason and justify the causes by using different statistical techniques. This study provides a broad overview of the time series data of covid confirmed cases and fatality rates globally and compares reported cases in India with other Countries computes the actual growth rate and daily changes for specific locations by estimating the global trends per location. Also, the SARS Cov-2 trend and forecasting for the next 28 days up from Dec 23. 2021 to Jan 19, 2022, is also predicted by applying different statistical tools and techniques

## **PROBLEM STATEMENT**

The novel coronavirus reported at the end of 2019 has impacted almost every aspect of life. This project mainly focuses on analyzing, forecasting, visualizing, and investigating the current coronavirus spread in different countries.

#### **OBJECTIVE**

- 1. To analyze the time-series data of covid confirmed cases and fatality rates globally
- 2. To compute the actual growth rate and daily changes for specific locations
- 3. To estimate the global trends per location
- 4. To compare reported cases in India with other Countries
- 5. To Predict the Pandemics Trend and forecast its data
- 6. To visualize the data by using Single Trend (for single location), itrend (interactive trends), mtrend (multiple locations)
- 7. To perform analysis for Vaccination & Testing and Genomics data

## **DATASET USED**

A Live data set package from R studio named "covid19.analytics," which fetches current covid data from the JHU Github repository, is used for the analysis.

## JHU's CCSE repository

https://github.com/mponce0/covid19.analytics

**Focused Attributes -** (Province\_State,Country\_Region,Confirmed, Deaths,Recovered and Active cases)

**Total Observation – 4006** 

**Total Variables** – 40

## TOOL DESCRIPTION(LIBRARIES, PACKAGES AND SOFTWARE USED)

Statistical Software used – R Studio

## Libraries Used:

- library(covid19.analytics)
- library(plotly)
- library(lubridate)
- library(ggplot2)
- library(prophet)
- library(dplyr)

# Techniques implemented:

• Summary Statistics

- ONE -WAY ANOVA
- Regression Analysis
- **SIR** MODEL (Susceptible-Infected-Recovered)
- Pandemics Trend Prediction and forecasting(time in day wise)
- Interactive visulaizations

#### CODE:

```
library(covid19.analytics)
library(plotly)
library(lubridate)
library(ggplot2)
library(prophet)
library(dplyr)
options(scipen = 6)
covidData <- covid19.data()</pre>
## obtain time series data for "confirmed" cases ##
confirmed_cases <- covid19.data(case = "ts-confirmed")</pre>
# India Confirmed Cases
confirmed_cases_in <-</pre>
 confirmed_cases %>% filter(Country.Region == "India")
print(confirmed_cases_in)
## Death Count / reads time series data for casualties ##
death_cases <- covid19.data(case = "ts-deaths")</pre>
# India Death Cases
death_cases_in <- death_cases %>% filter(Country.Region == "India")
print(death_cases_in)
## obtain time series data for "recovered" cases ##.
recovered_cases <- covid19.data(case = "ts-recovered")</pre>
# India Recovered Cases
recovered_cases_in <-</pre>
 recovered_cases %>% filter(Country.Region == "India")
print(recovered_cases)
```

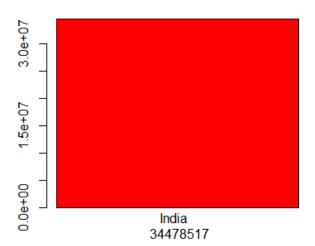
```
#Summary Statistics
#GLOBAL
report.summary(Nentries = 5,
          graphical.output = T)
report.summary(Nentries = 5,
         graphical.output = T,
          geo.loc = "India")
#one way anova
one.way = aov(Confirmed ~ Deaths, data = covidData)
summary(one.way)
# Graphs and Visualization
total_ts <- covid19.data(case = "ts-ALL")</pre>
totals.plt(total_ts)
#totals per location //Regression Analysis
tots.per.location(confirmed_cases, geo.loc = "India")
# growth rates
growth.rate(confirmed_cases, geo.loc = c('US', 'India'))
growth.rate(confirmed_cases, geo.loc = c('Brazil', 'India'))
#SIR modelling
SIR = generate.SIR.model(confirmed_cases, 'India', tot.population = 34478517)
#livemap
live.map(confirmed_cases)
# Pandemics Trend Prediction (time in day wise)
tsc <- covid19.data(case = 'ts-confirmed')</pre>
tsc <- tsc %>% filter(Country.Region == 'India')
tsc <- data.frame(t(tsc))</pre>
tsc <- cbind(rownames(tsc), data.frame(tsc, row.names = NULL))</pre>
colnames(tsc) <- c('Date', 'Confirmed')</pre>
tsc \leftarrow tsc[-c(1:4),]
tsc$Date <- ymd(tsc$Date)</pre>
str(tsc)
tsc$Confirmed <- as.numeric(tsc$Confirmed)</pre>
#Plot
gplot(Date, Confirmed, data = tsc, main = 'Covid19 confirmed cases in India')
```

```
ds <- tsc$Date
y <- tsc$Confirmed
df <- data.frame(ds, y)</pre>
# Forecasting
m <- prophet(df)</pre>
# Prediction
future <- make_future_dataframe(m, periods = 28)</pre>
forecast <- predict(m, future)</pre>
#Plot forecast
plot(m, forecast)
dyplot.prophet(m, forecast)
#Forecast components
prophet_plot_components(m, forecast)
#Model Performance
pred <- forecast$yhat[1:121]</pre>
actual <- m$history$y</pre>
plot(actual, pred)
#Single Trend for India
indiaData <-
 confirmed_cases[confirmed_cases$Country.Region == "India" , ]
single.trend(indiaData)
#Multiple Location
mtrends (confirmed_cases, geo.loc = c ("US", "India"))
#Interactive Locations trend daily cases
itrends (covid19.data("ts-confirmed") , geo.loc = "India")
itrends (covid19.data("ts-confirmed") , geo.loc = "ALL")
#TESTING
covidTest <- covid19.testing.data(tgt = "testing", disclaimer = TRUE)</pre>
#covidTest<-covidTest %>% filter(Country.Region == "India")
print(covidTest)
#VACCINATION
covidvaccine <-
```

## **OUTPUT**

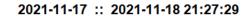
## TS-CONFIRMED Cases -- Data dated:

India 34478517

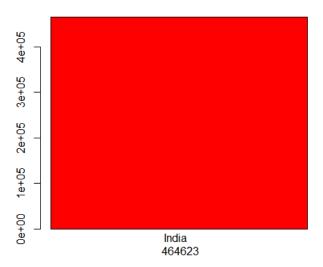


2021-11-17 :: 2021-11-18 21:27:08

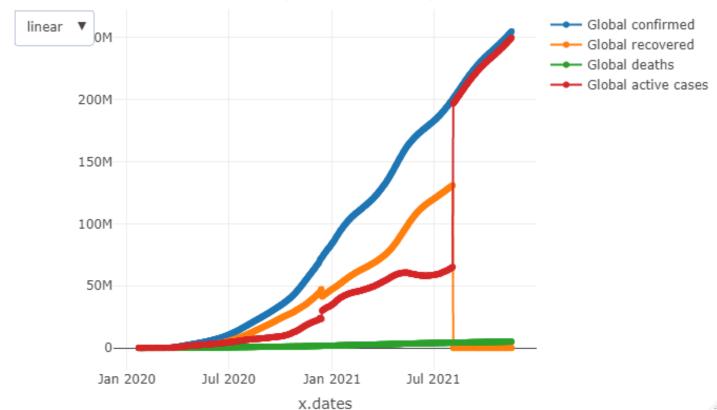
#### TS-DEATHS Cases -- Data dated: 2

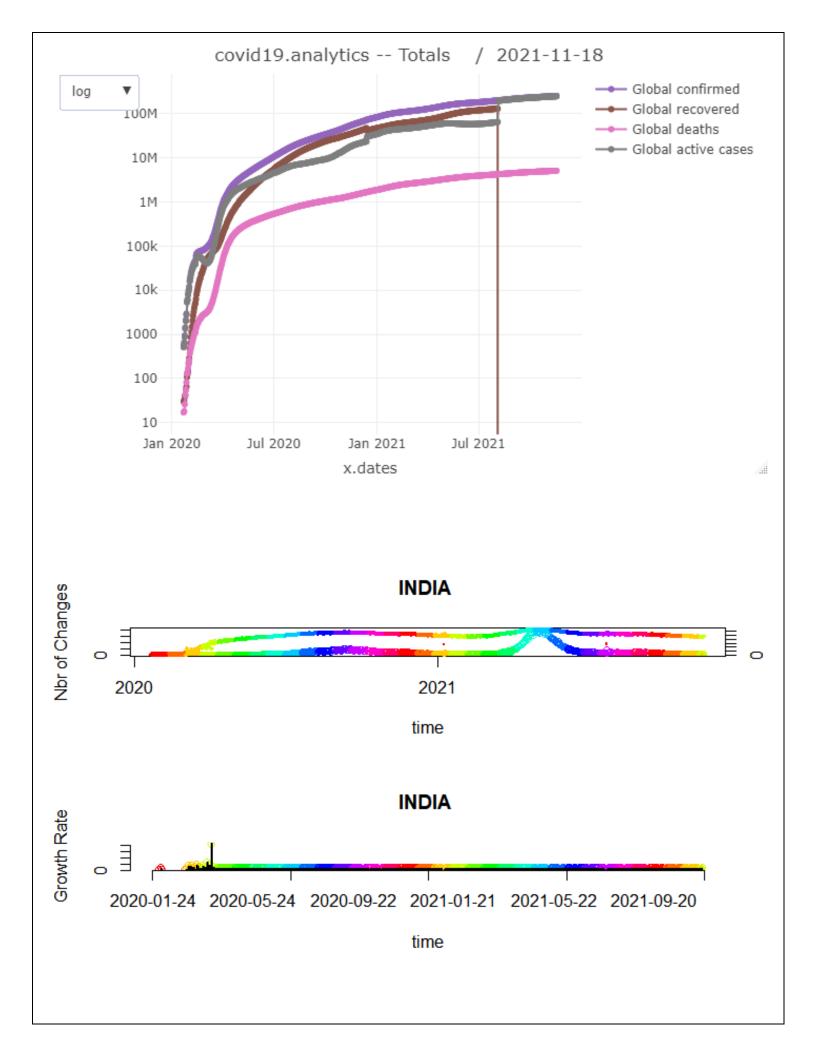


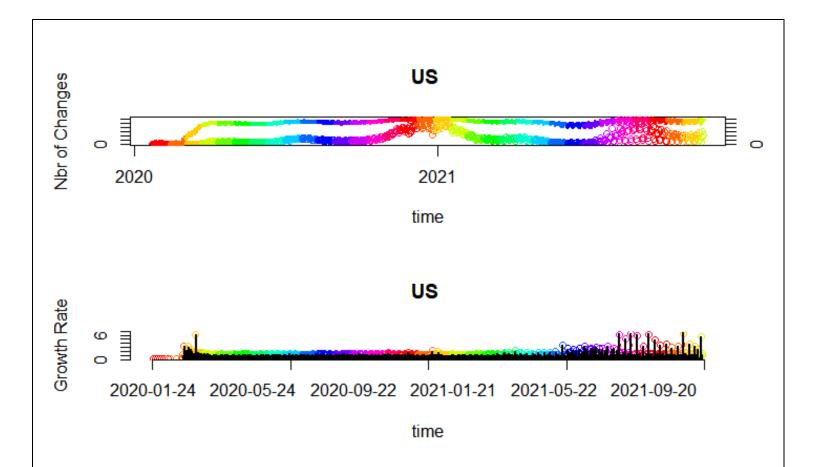


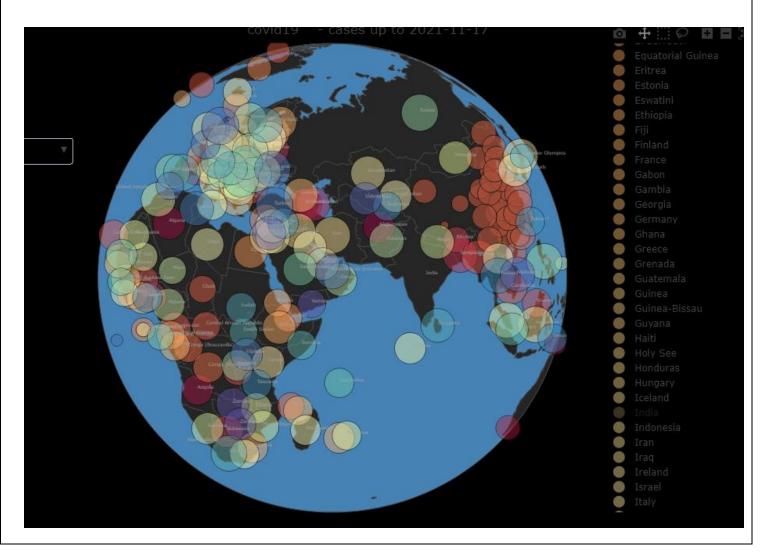


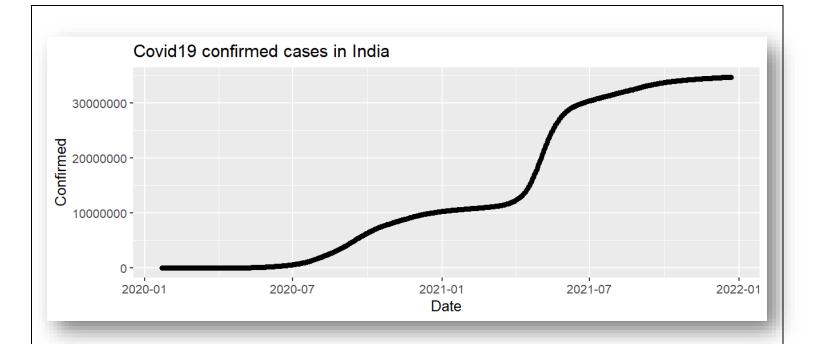


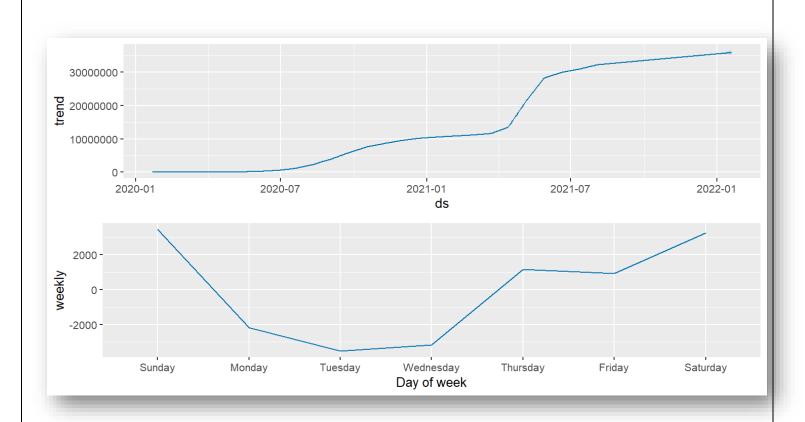


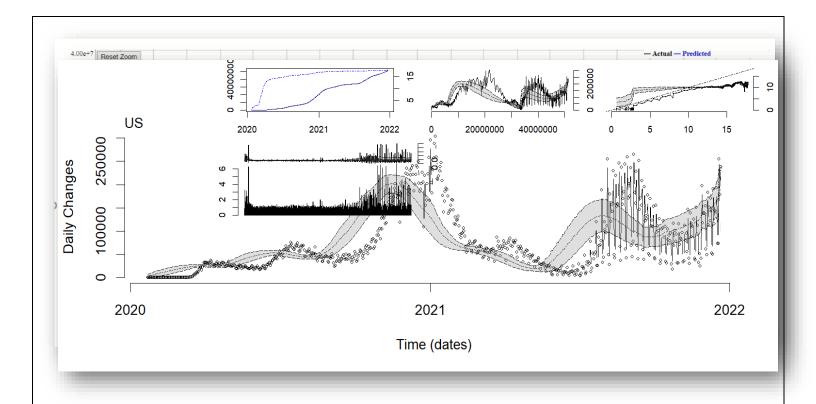


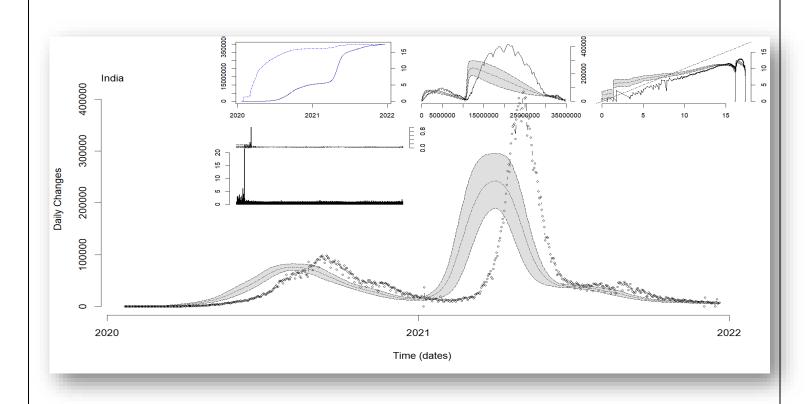


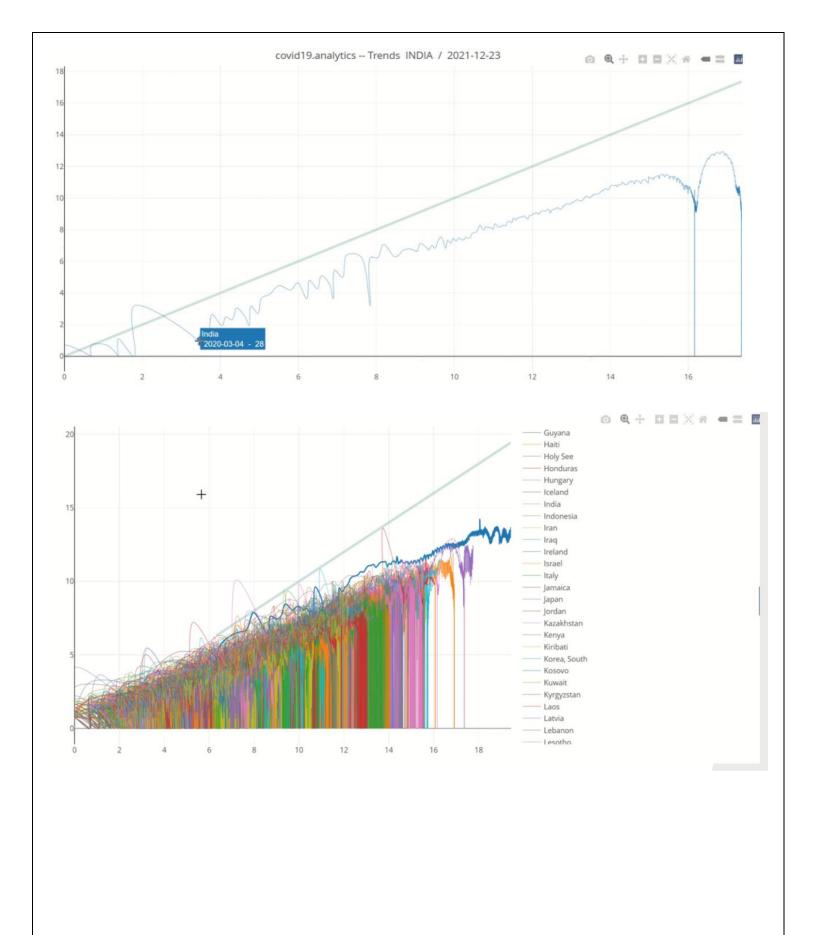


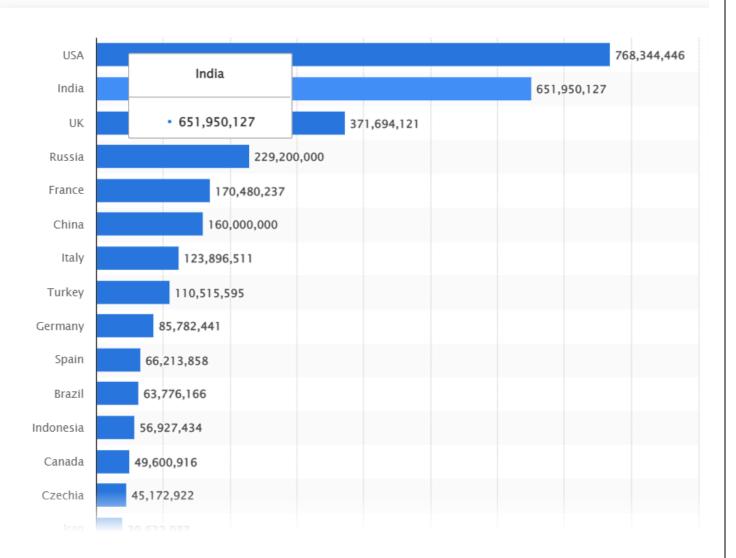


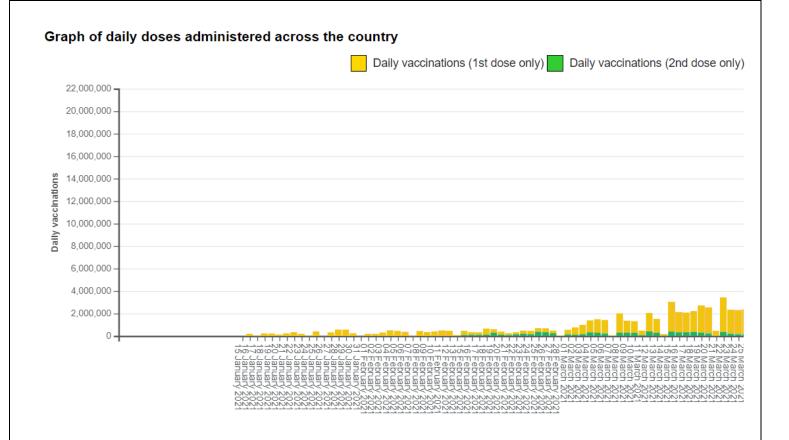




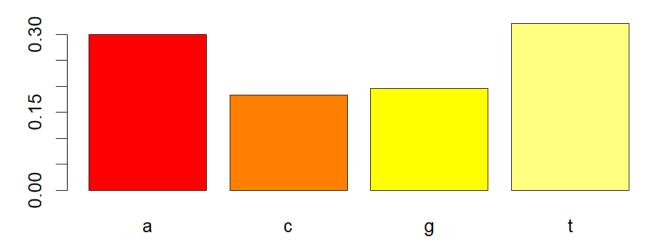








# ACTG Distribution in covid19 genome



## **ANALYSIS REPORTS AND INFERENCE:**

- Comparing Confirmed and Death cases globally shows that the two entities are significantly different as the F value is positive. We reject the Null Hypothesis. Infers that confirmed and instances of death vary, indicating that the confirmed cases are more than the death cases.
- Compared to India, brazil has more Covid cases
- Thus with the SIR model, the covid infected cases are getting increased in India as the number of persons who are prone covid is more
- Forecasting for the next 28 days predicts that the covid cases will increase concerning the actual data (up to Jan 19, 2022).
- It infers that the SARS COV-2 gets evolved in different variants by mutating its genome in a way by changing its characteristics and traits to spread even more vigorously, thus resulting in the rise of COVID cases globally

## PROOF for TRUTHFULNESS OF THE FORECASTING DATA

######################################	Data dated ########## s reported: reported: 8	d: 2021-11- ############# 196 38	-17 :: 2021- #############
Worldwide ts-confirmed To	tals: 2550043	381	
Country.Region Province.St	ate Totals	GlobalPerc	LastDayChange
L US	47420114	18.60	111106
l US 2 India 3 Brazil	34478517	13.52	11919
3 Brazil	21977661	8.62	11977
4 United Kingdom	9675058	3.79	37868
S Russia	9027163	3.54	35415
	6 (sd: 1.54)		

Worldwide ts-confirmed Totals: 277161199					
	Country.Region Province.State	Totals	GlobalPerc	LastDayChange	
1	US	51545991	18.60	238378	
2	India	34765976	12.54	7495	
3	Brazil	22222928	8.02	3451	
4	United Kingdom	11647473	4.20	105336	
5	Russia	10114983	3.65	25038	

23-12-2021

## **CONCLUSION:**

## JUSTIFICATION:

Though Herd immunity is developing among the people to fight against the novel coronavirus, it mutates itself by changing its characteristics and traits, resulting in a widespread virus.

## SOLUTION TO THIS PROBLEM:

- The sequencing of the Genomes and testing can be done to prevent the spread of the mutated virus
- The initiation of vaccination can increase among the people
- People can maintain the Covid restrictions put forth by the Government to bring the covid in control