Analysis and visualization of the pandemics using Artificial Intelligence

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Abstract. Human race has faced many epidemics and pandemics in past. The trajectory reveals that there is a pandemic almost every century. Our generation has witnessed the outbreak of coronavirus disease (COVID-19) pandemic, which turned out to be largest pandemic ever. Viruses have affected global population in the past century can answer the questions of the post Corona trauma. It has been noticed that most of the epidemics and pandemics are inflicted by the influenza viruses. Scientists have elucidated the structures of the viruses that have infected earlier. The homology between them and COVID-19 can be studied in order to fight the virus. In this paper, journey of novel corona virus has been discussed from a simple flu to become a pandemic within a short span of time. All the factors including the major countries responsible for spreading the pandemic owing to their volume, speed of human travel and socio-economic reasons seems to be the reason according to the review. We also compared the structures of the epidemic and pandemic causing viruses. Some of the COVID-19 studies mentioned which used machine learning models, and then by the use of machine learning we analyzed characteristics of three diseases (H1N1, COVID-19, and SARs) to obtain the plots of the autocorrelation against the lag of the number of confirmed cases, recovered cases, and deaths of /caused by COVID-19.

1. Introduction
The 21st century has registered 5 epidemics out of which 3 were caused by Coronaviruses. Coronavirus is a family of viruses which generally cause respiratory diseases or infect the upper respiratory tract [1]. Severe Acute Respiratory syndrome coronavirus (SARS-CoV) in 2003, Middle East Respiratory
syndrome coronavirus (MERS-CoV) since 2012 and the largest pandemic outbreak SARS-CoV2 which causes COVID-19 pandemic are all caused by viruses from Coronavirus family. Coronavirus diseases are caused by infection of virus in the respiratory tract. Like SARS-CoV-2, SARS-CoV was epicentered in China and infected approximately 8000 people worldwide causing about 744 deaths [2][28]. On the other hand, MERS is suspected to have originated from Saudi Arabia, where its first case was reported in 2012. Since then it has infected about 2494 people in 27 different countries causing 858 deaths globally. The rate of infection for SARS-CoV and MERS was relatively very low as compared to COVID-19. COVID-19 has so far infected 218 countries, with approximately 6.42 million people already infected, causing 3,83,000 deaths depicting how dangerous this new coronavirus is [3][19]. It has a very high reproductive rate and in extreme cases, it is reported to pass on from patient to healthy individual without any symptoms after infection, hence becoming more virulent and fatal.

Combating pandemics and epidemics has been an endless journey for the mankind. Genetic selection made viruses vulnerable to evolution. Humankind had witnessed epidemics like Bubonic plague,Ebola, Cholera, Encephalitis and many more (Organization; Patterson & Pyle, 1991; Who, 2011; Ying [4][20]. However, they have been successfully eliminated after development of vaccines, and necessary preventive measures. Since Black Death and Spanish flu, health care has advanced to achieve greater levels and in contrary, viruses and bacteria have evolved too. Hence, when all the pandemics and epidemics from last century were compared by taking into account the preventive measures, treatments, socio-economic factors etc., a key line can be drawn on efficient handling of coronavirus pandemic. Not only the research principles but also statistics and socio-economic factors of all the former tracks were considered to unfold the fate of the novel viruses causing an epidemic or pandemic.

1.1 Epidemics and Pandemics in past After summarizing epidemics and pandemics, authors depict that influenza viruses are the one which resulted in major wipeouts of population in compared to other viruses and bacteria. The seasonal flu had also been a pandemic taking nearly one million lives. Since the outbreak of Spanish flu, which caused the largest epidemic in the last century, to the COVID-19, all of them have been the influenza viruses that were responsible for annihilation across the world. Before the Coronavirus Pandemic, there have been viruses related to respiratory afflictions taking part in epidemics and pandemics such as Spanish flu, Asian flu, Hong Kong flu, London flu, Severe Acute Respiratory Syndrome (SARS), Swine flu, Middle East Respiratory Syndrome (MERS) (Vonk & Shackelford, 2012). Post Spanish flu trauma all the other viruses have been controlled after the onset (Friedson, McNichols, Sabia, & Dave, 2020). The genome sequences of SARS, MERS-CoV, and COVID-19 have been found out to be homologous (Patterson & Pyle, 1991).

| Pandemic/Epidemic | Virus | Envelope | Capsid Structure | NUCLEIC ACID |
|-------------------|-------|----------|-----------------|--------------|
| Spanish flu (1918-20) | H1N1 | Present | Helical nucleocapsid | Single-stranded negative sense RNA strand |
| Asian flu (1957) | H2N2 | Present | Helical nucleocapsid | Single-stranded negative sense RNA strand |
| Hong Kong (1968) | H3N2 | Present | Helical nucleocapsid | Single-stranded negative sense RNA strand |
| Small pox | Variola virus | Present | Brick shaped with complex internal structure | Double stranded linear DNA |
| Polio | Enterovirus | Absent | Icosahedral shape | Positive sense RNA virus |
| HIV | Lentivirus (retrovirus) | Present | Conical | Positive sense RNA virus |
| Ebola | Zaire Ebola Virus | Present | Helical nucleocapsid | Single stranded negative sense RNA |
| Corona virus | SARS-CoV | Present | Helical nucleocapsid | Positive sense RNA virus |
Novel corona virus COVID-19 Present Spherical or pleomorphic Positive sense RNA virus

Spanish flu trauma all the other viruses have been controlled after the onset. [5] The genome sequences of SARS, MERS-CoV, and COVID-19 have been found out to be homologous. [6]

Several countries have been affected due to these pandemics and epidemics. Authors found that the most devastated ones by the fatality of the epidemics are USA, New Zealand, United Kingdom, European region, South Africa, Hong Kong, Nigeria, the Democratic Republic of Congo as in Figure 2 [7]. Besides, the soaring mortality rate can be seen in the 20th century when compared to 21st century as in Figure 3 and 4. Due to the pandemic outbreaks such as Bubonic Plague, Spanish flu and Encephalitis in the early 20th century, the fatality rates have been greater whereas in the late 20th century because of the Small Pox taking 500 Million lives and more 50 million lives by Spanish flu as in Figure 3 [8].

Figure 1. (a) Time line of pandemic and epidemics and the infecting viruses. This gives us hope that COVID-19 can be taken care of as other deadly viruses had been and (b) a Phylogenetic tree to show the origin and similarity of virus from different times as listed in Table (1).
In 21st century, outbreak of Swine flu has caused 575,400 deaths across the world and more significantly in the USA, Northern Ireland, India, Nepal, Pakistan and middle Asian Countries as in Figure 4. Seasonal Influenza in the late 20th century witnessed about 80,000 cases of mortality to become the more severe pandemic of the century as in Figure 5[9][18]. Thus, an easy conclusion can be drawn that USA, USSR, South Africa and East Asian countries are more vulnerable to pandemics and epidemics[10][24]. The mortality rate (Figure 4) does not depend only on the disease inflicted by the virus but also on the socio-economic reasons that do not allow one to take enough measures from being exposed to the virus[11][35].

1.2 Viruses and Pandemics  Viruses have inherent advantage in preserving and transmission of the information because of their relatively simple structure compared to other prokaryotic and eukaryotic organisms [12][36]. When dealing with a virus, one needs to keep in mind the structure of the virus as understanding them can tell us how to prepare drugs that can be used to fight them [13][37].
Table 2: Bibliographic summary of similar work already published

| References | Objective | Methodology Used | Conclusion |
|------------|-----------|------------------|------------|
| [27]       | To help doctors identify COVID-19 patients based on early symptoms and to separate out severe cases. | A prognostic prediction model based on XGBoost machine learning algorithm. | The 3 indices-based prognostic prediction model predicted the mortality risk, and present a clinical route to the recognition of critical cases from severe cases. |
| [19]       | This analysis was aimed at tracing a trend related to death counts expected at the 5th and 6th week of the COVID-19 in India. | Multiple regression and linear regression analyses. | According to the analysis, projected death rate (n) is 211 and 467 at the end of the 5th and 6th week from now, respectively. |
| [30]       | To develop a fully automatic framework to detect COVID-19 using chest CT and evaluate its performances. | A deep learning model and neural network (COVNet) for COVID-19 detection. | Identify COVID-19 on chest CT exams & community acquired pneumonia on chest CT exams. |
| [31]       | To estimate the number of COVID-19 infected patients by April-2020 and the required healthcare system to cope with it. | Data collected from reliable sources for prediction of possible number of cases in coming days. | If the exponential trend continues, more than 2500 hospital beds in intensive care units will be required within a week in Italy. |
| [32]       | To review the basic reproduction number ($R_0$) of the COVID-19 virus. | PubMed, bioRxiv and Google Scholar were used to search for eligible data. | Estimated mean $R_0$ for COVID-19 is around 3.28, with a median of 2.79 and IQR of 1.16, which is considerably higher than the WHO’s estimate at 1.95. |

2. Methodology. In the present scenario, drugs in the form of monoclonal antibodies have shown progress against previous viral outbreaks like SARS-CoV and MERS-CoV but they haven’t been much effective against the new Novel coronavirus COVID-19.[14][18] This suggests that there are a variability of epitopes involved in the monoclonal antibodies.[15][21] Studies have shown that they have a great tendency for gene recombination, and adaptive mutation. Thus, focusing on its capability to attach to host receptors for vaccine production directly won’t solve the problem as virus evolves and enhances its affinity for various other receptors present on host cells.[16][20]

Since the plague, people were aware of quarantine and isolation as the disease spread via rats and fleas, where the infection could not be prevented. Thus, limited movement and transport patterns prevented the spread of infection.[17][23].

The extent of the pandemic successively reflected the increasing expansion of global transport and movements. The speed of spread left the world population susceptible to infection. The availability of regular air and seaway traffic is one of the major cause for the spread of novel coronavirus. The fatality rate for the pandemic is less as compared to other pandemics but the speed and volume of spread are many folds higher than the others, indicating its high risk of increased global transport and further spread of other pandemics.

In this paper, we have studied the journey of spread of COVID-19 and how it became a pandemic. China is the epicenter for spread of the novel coronavirus and the outbreak reached all through USA, South Korea, Iran, middle-east region, European countries, Turkey significantly. The novel coronavirus has high mortality rates in countries known for people travelling around the world as given the Table 1 and Table 2 obtained from WHO situation reports. As of June 8, 2020, out of 197 countries in the world that
have been affected by COVID-19, except Kiribati, Marshall Islands, Micronesia, Nauru, North Korea, Palau, Samoa, Solomon Islands, Tonga, Turkmenistan, Tuvalu, and Vanuatu.

Table 3. Approximate number of people infected dead and recovered due to COVID-19. (Until 30/05/2020)

| Country   | Total Infections | Mortality Cases | Cured Cases |
|-----------|------------------|-----------------|-------------|
| Australia | 7171             | 103             | 6582        |
| Brazil    | 468338           | 27944           | 189476      |
| Canada    | 89418            | 6979            | 47518       |
| China     | 83000            | 4634            | 78302       |
| Egypt     | 22082            | 879             | 5511        |
| France    | 149668           | 28714           | 67803       |
| Germany   | 183019           | 8594            | 163941      |
| India     | 173763           | 4971            | 82370       |
| Italy     | 232248           | 33229           | 152844      |
| Japan     | 16759            | 882             | 14147       |
| Netherlands | 46126         | 5931            | -           |
| Poland    | 23155            | 1051            | 10692       |
| Portugal  | 31946            | 1383            | 18691       |
| Russia    | 387623           | 4374            | 159257      |
| Saudi Arabia | 81766       | 458             | 57013       |
| South Africa | 27403       | 577             | 14370       |
| South Korea | 11441         | 269             | 10398       |
| Spain     | 238564           | 27121           | 150376      |
| United Kingdom | 271222   | 38161           | -           |
| United States | 1783132    | 104166          | 384821      |

Table 4. Approximate number of people infected dead and recovered due to SARS, H1N1 and COVID-19. (Until 30/05/2020)

| Country    | Disease | Confirmed Cases | Mortality | Cured Cases |
|------------|---------|-----------------|-----------|-------------|
| China      | SARS    | 5327            | 349       | 4949        |
| Canada     | SARS    | 251             | 41        | 200         |
| Singapore  | SARS    | 238             | 33        | 205         |
| Viet Nam   | SARS    | 63              | 5         | 58          |
Table 5. Approximate number of infections, deaths caused by COVID-19 and number of recovered cases. (Until 30/05/2020)

| Continent         | Total Cases | Mortality | Cured Cases |
|-------------------|-------------|-----------|-------------|
| Asia              | 1085667     | 29756     | 653460      |
| Africa            | 137677      | 3945      | 58225       |
| Europe            | 1996536     | 172742    | 973249      |
| North America     | 2015861     | 122348    | 649802      |
| Oceania           | 8793        | 125       | 8187        |
| South America     | 800398      | 38185     | 327866      |

Figure 3. Overview of methodology.
For this study, global pandemic data was collected from WHO and worldometer for consecutive date in the .xlsx format which was converted to .csv format for formatting, training, validation and testing. The CSV data file was then uploaded on Jupyter notebook for analysis with python 3.7 (64-bit Graphical Installer). The Input Values included corresponding column names from the dataset.

The collected dataset was unstructured data with several null or unstructured values which cannot undergo pre-processing. This unstructured data was structured by removing all the null values which were not appropriate for the study or may cause bias and errors while execution of the codes. Required Python Libraries were imported into the Jupyter Notebook to provide predetermined and pretested set of helper functions for specific functionality. The format of dataset used in this study is summarized in table 3, 4, and 5.

3. Results. Since its appearance in the city of Wuhan (Hebei district) in China, Coronavirus has been a relentless march of new cases and deaths [22][28]. WHO along with governments have already taken strong measures including complete or partial lockdown of both its internal and external borders to restrict the mobility along with social isolation of the suspected and confirmed patients. If the situation continues in the present state; projected death rate (n) is 211 and 467 at the end of the 5th and 6th weeks from now, respectively [29]. R0 estimates for SARS have been reported to range between 2 and 5, which is within the range of the mean R0 for COVID-19 and due to similarities of both pathogen and region of exposure, this is expected but, despite the heightened public awareness and the impressive strong interventional response, the COVID-19 is already more widespread than SARS, indicating it may be more transmissible [23]. Due to high transmission rate COVID-19 has now become a pandemic that has led to shut down to the world’s most powerful nations like the USA, France, Italy, etc. It is affecting the world economy, stocks, share markets etc. The poorer nations with less developed medical facilities are worst affected. In India, daily wage workers are suffering from the loss of income which has led them to step out on their foot, with their families in the hope to reach their hometown amid lockdown. The infection is rising under these circumstances. In our study we have collected the data of three pandemics of 21st century: SARS-COVID which spread in 2003, H1N1, or swine flu which spread in 2009-10 and COVID-19. We have used heat maps to compare the data. Heat maps and clustering are used frequently in expression analysis studies for data visualization and quality control [26][33][38]. Heat maps use color gradient for analysis of data which makes it possible visualize the huge volume of data easily and effectively. In this study, we used heat maps for analysis of following characteristics of three diseases (H1N1, COVID-19, and SARS): no. of suspected cases, no. of probable cases, no. of confirmed cases, no. of suspected deaths, no. of probable deaths, no. of suspected deaths. k-Nearest-Neighbors(k-NN) is implemented further as a supervised learning model, as the model here learns from the input data and makes out the predicted value that isn’t seen. As the data is split into the input value and the target value, the building and the training of the data and for training the model, fit function was used and once our data was trained the predictions was be made. In our work, the value of the K is determined by the optimal classification or the accuracy classification. The value is determined to be 11 with the maximum depth of 3 and n_neighbors to be 5 with a range of (1, 15) and an accuracy of 79.41176470588235 was achieved. Fig 4(h) plots the accuracy (y-label) and the k values (x-label), the values for y were disease which were the target value. The auto co-relation with the lag zero was always equal to 1 because it represents the autocorrelation between each term and itself. It was calculated using the following equation [25] [39].

\[
\hat{\gamma}(h) = \frac{1}{n} \sum_{t=1}^{n-|h|} (x_{t+h} - \bar{x})(x_t - \bar{x})
\]

Value and value with lag zero will always will be the same.
4. Conclusion. COVID-19 spread the entire globe within a matter of months. One of the main reasons for the quick spread of the disease is the volume and speed of human travel. The traveler across the globe were unaware of the infection they were carrying and unknowingly spread it to several countries. Secondly, the capacity of the virus to survive in the different weather and spread quickly through cough and cold is another reason for quick spread. Third, the irresponsible behavior of the people in quarantine added to the spread of the virus. Lastly, the unavailability of the vaccine or medicine to combat the virus makes it difficult for the doctors to fight the disease. In this study, the epidemics and pandemics that occurred in the last century, the number of countries they affected, and their mortality rates were discussed so as to compare them. Then the structures of the viruses that caused the pandemics were discussed to understand the pathogenicity. And then the reasons why tackling Covid-19 is harder to deal with when compared to other viruses are discussed along with the number of countries that are affected by Covid-19, the mortality rate. Then some COVID-19 studies that used the artificial intelligence/machine learning models were discussed. Then by the use of one such model, analysis of following characteristics of three diseases (H1N1, COVID-19, and SARS): number of suspected cases, number of probable cases, number of confirmed cases, number of suspected deaths, number of probable deaths, and the number of suspected deaths, was done. It was seen that the amount of confirmed cases of COVID-19, amount of deaths caused by COVID-19 and the number of recovered patients recovering from COVID-19 are more when compared to the same numbers of H1N1 and SARS. Also, a heat map was prepared to show the correlation between the number of confirmed cases, number of deaths, and the number of recovered cases of H1N1, COVID-19, and SARS. Also, three plots representing the autocorrelation against the lag of the number of confirmed cases, the number of deaths caused by COVID-19 were prepared. The lockdown and social distancing in many countries have helped overcome the fast spread of the virus, however, this has resulted in slowing down of the economy. The only blessing in disguise is that Mother Nature got a chance to heal itself. In conclusion, we can interpret that it is not the first time that a pandemic has hit the globe. Earth has seen epidemics and pandemics other than COVID-19. Scientists, pharmaceuticals, and global leaders have always come up with ways to deal with epidemics and pandemics. The scientists and pharmaceutical companies across the globe have put up some commendable work on a war footing in the field of developing testing kits, developing vaccines and preparing post-symptomatic treatment.
protocols to deal with COVID-19. Social distancing and lockdowns in many countries have already reduced the rate of spread and the hope is the vaccine will be available sooner than later.

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