

Review On Covid Data Analysis Using Spatial Temporal Data

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Abstract

Coronavirus (Covid) is a severe acute respiratory syndrome infectious disease, spreads mainly between human beings during close contact, most symptoms are coughing, sneezing, and speaking small droplets. The outbreak of COVID has led to a worldwide socio-economic crisis, with major impacts on developing countries. Understanding the dynamics of the disease and its driving factors, on a small spatial scale, might support strategies to control infections. It explores the impact of the COVID-19 on neighborhoods of Recife, Brazil, for which we examine a set of drivers that combines socio-economic factors and the presence of non-stop services. A three-stage methodology was conducted by conducting a statistical and spatial analysis, including clusters and regression models. COVID-19 data were investigated concerning time dates between April and July 2020. Hotspots of the most affected regions and their determinant effects were highlighted. We have identified that clusters of confirmed cases were carried from a well-developed neighborhood to socially deprived areas, along with the emergence of hotspots of the case-fatality rate. The influence of age-groups, income, level of education, and the access to essential services on the spread of COVID-19 was also verified. The recognition of variables that influence the spatial spread of the disease becomes vital for pinpointing the most vulnerable areas. Consequently, specific prevention actions can be developed for these places, especially in heterogeneous cities.

Keywords

Covid-19, Spatial-Temporal analysis, Variable Importance analysis, Pandemic, Medical Risk Factors, Neighborhood

Introduction

The World Health Organization (WHO) declared the coronavirus disease 2019 (COVID-19) pandemic in March 2020. It is a viral disease caused by the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) virus. In the early stages of the pandemic, the spread of the disease was primarily reduced by non-pharmaceutical interventions, which were then complemented with vaccination programs as technology evolved. However, changes of intervention measures along with the emergence of highly infectious variants of the virus have affected the rates and the patterns of the spread of

COVID-19. Countries are grappling with many tactics in order to minimize the spread of Covid: collection, close schools, stop transportation, lock towns, enforce curfew, and seal places, but unable to contain it effectively. The time is required to locate the risk assessment on a site basis to take prompt preventive measures. Globally, there are 48,93,195 cases of coronavirus, while the death toll is 3,22,861. Taking into account the updates of the Ministry of Health on Wednesday (May 20, 2020), India received a total of 1,06,750 COVID cases which include 61,149 active cases, 42,298 cure and 3303 deaths [3]. In the last 24h, there have been 5611 new cases and 140 deaths. The rate of recovery is 39.62%.

Spatial-temporal analysis might be a valuable tool to capture the dynamic behavior of the pandemic. One of the most regularly applied approaches in this context is spatial analysis, in which cluster and hotspot analysis, interpolation, and space–time scan statistics have been the main techniques used. For instance, in China, spatial clustering patterns, overtime, of cumulative cases of COVID-19 at the city and county-levels showed a shift from hotspots of larger coverage to few specific places. In global terms, evidence suggests that there is a sort of spread of COVID-19 cases from developed countries to less affluent countries.

South America saw its geographic centroid for COVID-19 cases generally shifting from west to east in Brazilian areas, up to January 2021. In Brazil, the number of cases and deaths started in São Paulo state (Southeast region), then progressively moved north until early May 2020. The COVID-19 epidemic increased rapidly across northeastern Brazil up to May 2020, at which time it was the metropolitan areas that were reported as having clusters of cases, even though there was an already relevant spread towards the countryside. Understanding the transmission dynamics within local communities is important in order to develop strategies to combat the spread of the disease.

METHODOLOGY

A 7-day moving average was applied to the data on daily infections at the city-level, thereby seeking to reduce sudden variability due to reporting biases such as the lack of testing and the delay in recording cases and deaths. The impact of restricting and relaxation measures imposed by the State Government of Pernambuco and the City Council of Recife on the variability of the rate of new infections was also discussed. These government actions include the following: closing of non-essential commercial activities; mandatory use of masks; strict quarantine; reopening of building supply stores, beauty salons, suburban retailers, malls, and places of worship. In order to comprehend disparities in the spread of COVID-19 on a small scale, neighborhoods were explored using a spatial cluster analysis as applied for other diseases. Georeferenced data obtained from ten specific days between 16 April and 3 July 2020 at intervals of roughly one week were considered. Nevertheless, we were able to represent the ascending, peaking and descending behaviors of the curve of infections.

Local Moran's I was implemented for cumulative confirmed cases and the case-fatality rate—which means the total number of deaths divided by the total number of cases. This statistic assessed the spatial autocorrelation associated with each neighborhood of the study region in terms of a few surrounding spatial units. It was calculated for each area $i = 1, \dots, n$ in:

$$I_i = \frac{\sum_{j=1}^n W_{ij}(y_i - \bar{y})(y_j - \bar{y})}{\frac{1}{n} \sum_{j=1}^n (y_j - \bar{y})^2}$$

where y means the COVID-19 reported cases or case-fatality rates for the i th area or its j th neighbor in areas, and W_{ij} means a weight that represents proximity for the pair of areas i and

In this method, a kernel function with a bandwidth parameter is used to calculate a local weights matrix in terms of the distance between each pair of spatial units. Considering $i = 1, \dots, n$ as each sample location, GWR is mathematically denoted by :

$$y_i = \beta_{i0} + \sum_{k=1}^{p-1} \beta_{ik} x_{ik} + \epsilon_i$$

(2)

where y_i is the dependent variable of COVID-19 confirmed cases at neighborhood i , x_{ik} is the value of the k th explanatory variable at location i , β_{i0} is the intercept, β_{ik} is the regression coefficient for the k th explanatory variable, p is the number of regression terms, and ϵ_i is the random error at location i .

Cumulative count data has been applied in studies of COVID-19, especially at a small-scale level such as neighborhood or county-level and grid .

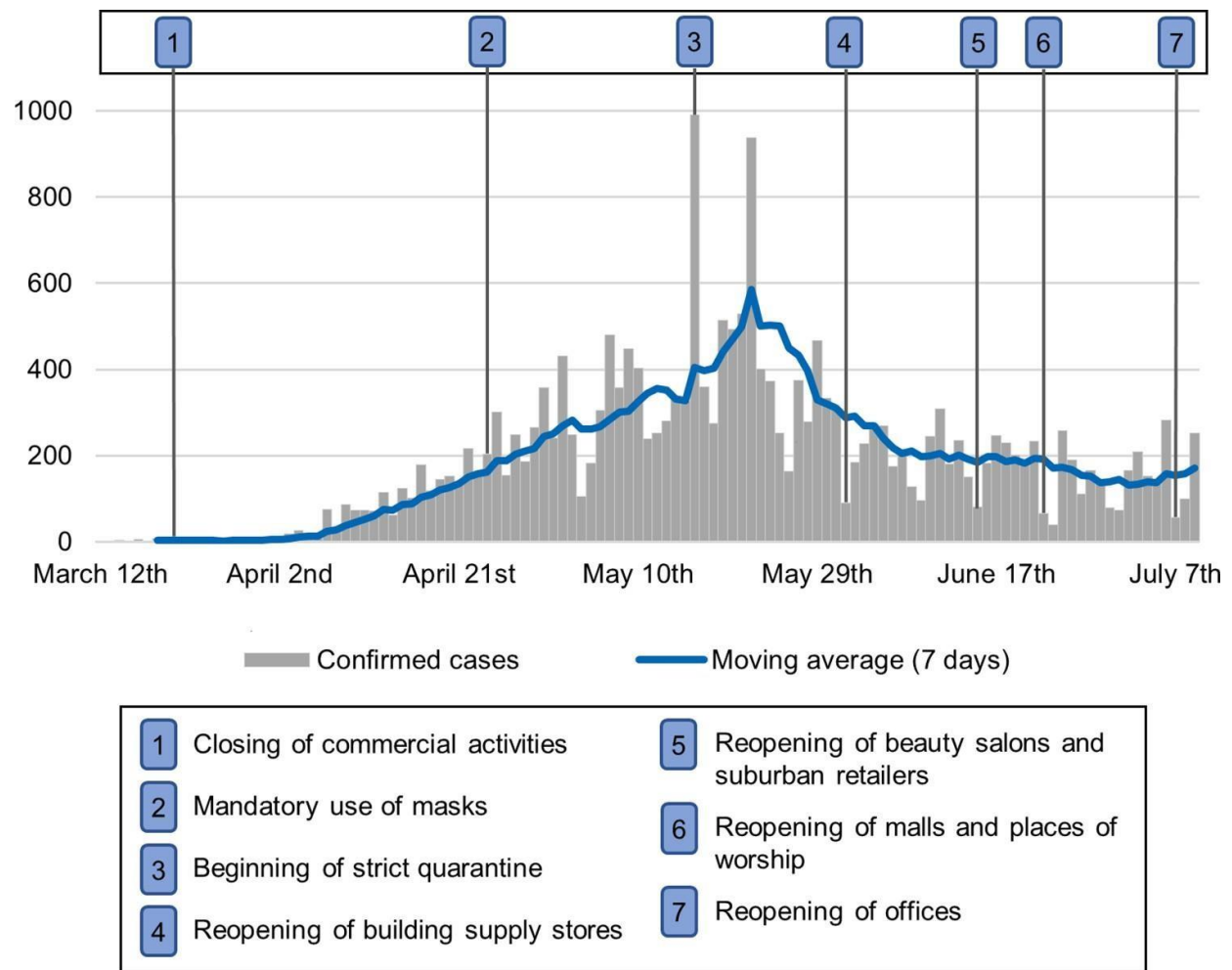


Fig1. Evolution of daily COVID-19 cases in Recife from March 12th to July 8th, 2020.

CONCLUSION

This study combined spatial clusters and statistical analysis to evaluate the influence of socio-economic factors and essential services on the spread of COVID-19 in the city of Recife, Brazil. Our findings reveal that an increased risk of transmissions was associated with children and the elderly, the size of the population, household income, the level of education, and the presence of some facilities that have remained open throughout the pandemic. Moreover, the spatial spread of the disease occurred by moving from well-developed to deprived neighborhoods during the initial stages of the pandemic. What was also found was for there to have been a tendency for there to have been harsh impacts on socially vulnerable and densely populated communities, specially those with many everyday places that are prone to overcrowding (e.g. bakeries, grocery stores). Brazil manages a public health system that is widespread in all federative units, even though the country has continental dimensions and faces complex challenges. However, currently, public agencies have been going through management difficulties. In this sense, this study can support strategic decisions to help mitigate the spread of COVID-19 not only in Brazil, but also in other developing and economically emerging countries. Furthermore, in the long term, knowledge produced during the COVID-19 pandemic in this heterogeneous context, regarding local characteristics and spatiotemporal patterns, can be used to structure policies for tackling new epidemics of viral infectious diseases.

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