Geo-Spatial analysis of SARS-CoV-2 infection in uMgungundlovu – Kwa-Zulu Natal, South Africa

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Outline of the Presentation

- Introduction
- Aim and Objectives
- Study setting
- Methods
- Results
- Discussion
- Conclusion
- Acknowledgements
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Introduction

- In 2019 a novel coronavirus disease was identified, epidemiologically linked to a wet market in Wuhan, China; virus :SARS-CoV-2; disease :COVID-19 [1,2].
- 1st September 2024, the World Health Organization recorded 776,137,815 cases and 7,061,330 deaths[3].
- January 2022, South Africa experienced four recognizable COVID-19 pandemic waves with 3-month periods of low transmission between each wave [5].
- Wave 1: ancestral strain with an Asp614Gly mutation, Wave 2: beta variant (B.1.351), Wave 3: delta variant (B.1.617.2), Wave 4: omicron variant (B.1.1.529) [5].
- Research needs include crossing different variables to interpret transmission including spatial analysis and spatiotemporal dimensions; geographical impact on public health measures; and predictive modelling of the evolution of the disease. Thus, using geospatial and statistical tools has become particularly relevant [6].

Aim and Objectives

Aim

The aim of this research project is to utilize geospatial analysis in understanding the distribution patterns of SARS-CoV-2 and its relationship with certain co-existing factors.

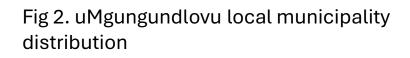
Specific objectives

- 1. Descriptive analysis of SARS-CoV-2 cases within uMgungundlovu District.
- 2. Investigate the spatial patterns of SARS-CoV-2 within uMgungundlovu District.

Study setting



Fig 1. KwaZulu-Natal District distribution



The Msunduzi

Richmond

Impend

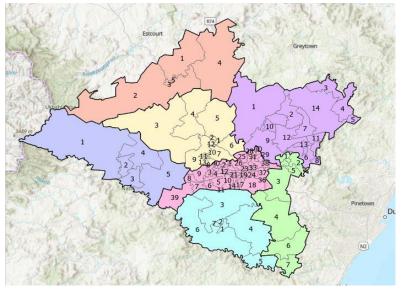


Fig 3. uMgungundlovu ward distribution

Methods

- All laboratory confirmed positive cases during the surge periods formed part of the study sample.
- Laboratory confirmed SARS-CoV-2 positive cases, including individual level data such as age, gender, ward number obtained from The National Institute for Communicable Diseases daily line lists.
- Statistical analyses were conducted using the R statistical software for data analysis and programming.
- Spatial characteristics of SARS-CoV-2 were investigated over the first four waves of transmission using ESRI ArcGIS Pro v2.0.
- Including Local Indicators of Spatial Association (LISA) with Moran's "I" as the measure of spatial autocorrelation; and Kernel Density Estimation (KDE).
- Base layers used to complete spatial analysis municipal demarcation board (open data source)

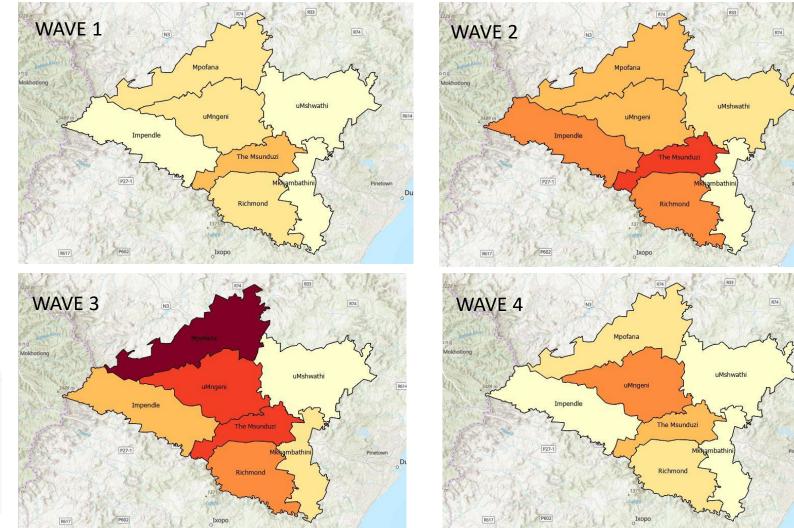
Results

Table 1: Descriptive analysis of SARS-CoV-2 cases

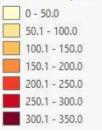
	Wave 1 – 10 304 cases	Wave 2 – 17 159 cases	Wave 3 – 20 765 cases	Wave 4 – 10 569 cases
Gender (incidence rate per 10 000)	Female: 112.1 (86.2 to 138.0) ¹ Male: 74.6 (47.7 to 101.5) ¹	Female: 180.8 (155.0 to 206.6) ¹ Male: 122.4 (95.6 to 149.1) ¹	Female: 207.0 (181.2 to 232.7) ¹ Male: 168.3 (141.5 to 195.0) ¹	Female: 107.8 (77.8 to 115.1) ¹ Male: 79.7 (52.9 to 106.6) ¹
Age	Mean age: Male: 40.1 years Female: 40.5 years (p = 0.58) ²	Mean age: Male: 41.55 years Female: 40.89 years (p = 0.28) ²	Mean age: Male: 33.66 years Female: 35.53 years (p<0.001) ²	Mean age: Male: 39.48 years Female: 38.82 years (p = 0.058) ²
Age group (incidence rate per 10 000)	Highest- 80 years&older: 262.4 (29.5 to 495.3) ¹ Lowest- 0-9 years: 14.2 (-25.1 to 53.5) ¹	Highest- 70-79 years: 350.4 (211.4 to 489.5) ¹ Lowest- 10-19 years: 64.9 (22.8 to 107.1) ¹	Highest- 40-49 years: 315.9 (257.1 to 374.6) ¹ Lowest- 0-9 years: 118.5 (79.4 to 157.5) ¹	Highest- 70-79 years: 214.9 (74.8 to 354.9) ¹ Lowest- 10-19 years: 44.1 (1.9 to 86.3) ¹
Municipality(i ncidence rate per 10 000)	Highest- Msunduzi: 137.4 (112.7 to 162.2) ¹ Lowest- uMshwathi: 24.7 (-35.3 to 84.7) ¹	Highest- Msunduzi – 206.4 (181.7 to 231.0) ¹ Lowest- Mkhambathini: 61.7 (-21.7 to 145.1) ¹	Highest- Mpofana – 304.8 (201.5 to 408.0) ¹ Lowest- uMshwathi – 55.8 (-4.1 to 115.8) ¹	Highest- uMngeni: 175.8 (112.0 to 239.6) ¹ Lowest- uMshwathi: 28.7 (- 31.3 to 88.7) ¹

 $10 \Gamma 0/O l ^{2}$ Fishers Event Test

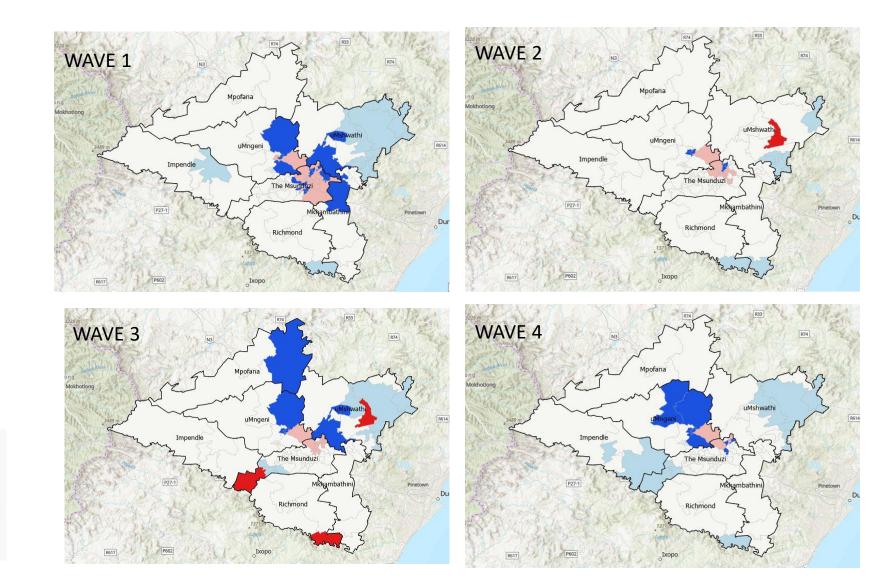
Choropleth map of SARS-CoV-2 incidence rate per LM, per wave



Incidence rate per 10 000 population

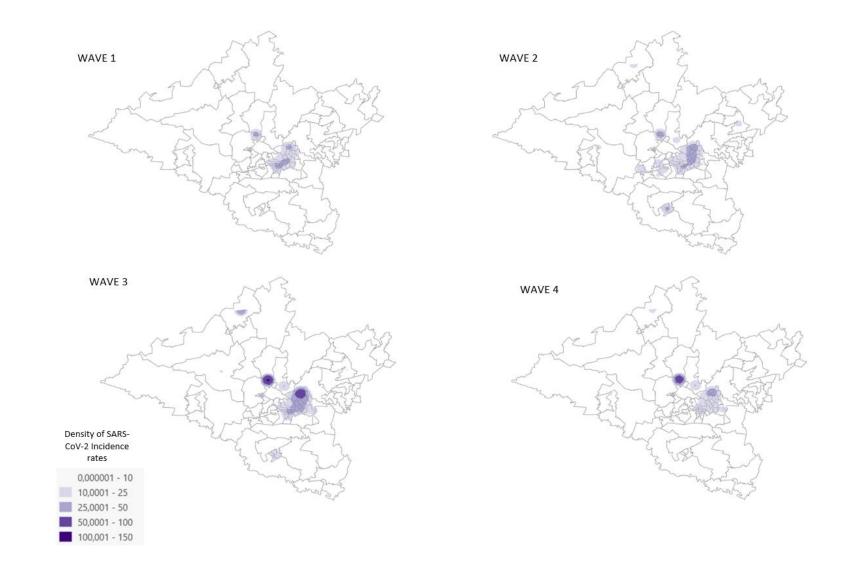


LISA analysis of SARS-CoV-2 incidence rate per wave



High-High cluster High-Low outlier Low-High outlier Low-Low cluster Not significant

SARS-CoV-2 incident case density maps per wave



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Discussion

- Higher incidence rates in females, urban areas, and older age groups. However, change in incidence rates over time.
- Under-reporting (stigma, fear of hospitalization, hard to reach areas)
- Predominant circulating variant per wave
- Vaccination coverage
- Health seeking behavior
- Immune related response
- Wave 1- Lockdown, fear of the unknown; Wave 2 super-spreader events; Wave 3 social unrest, congregate setting; Wave 4 – quick, omicron

Conclusion

- Relationship between infection and increased economic activity, population density and urban development, driven by increased contact.
- Differences in infection across waves highlight important characteristics of the predominant circulating variant, providing useful information on the evolution of the virus.
- The nature configuration of the social and built environment may be associated with increased infection.
- A Geospatial approach of analyzing infectious disease transmission is proposed to guide control efforts (e.g., testing/tracing and vaccine rollout).
- However, locally specific empirical research is required to assess other relevant factors associated with increased infection.

Acknowledgements



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References

- Schuchat A, & CDC COVID-19 Response Team. Public Health Response to the Initiation and Spread of Pandemic COVID-19 in the United States, February 24-April 21, 2020.
 Morbidity and mortality weekly report 2020; 69 (18):551-556.
- 2. Yang L, Liu S, Liu J, Zhang Z, Wan X, Huang B, Chen Y, Zhang Y. COVID-19: immunopathogenesis and Immunotherapeutic. Signal Transduction and Targeted Therapy 2020; 5:128.
- 3. WHO. WHO Coronavirus (COVID-19) Dashboard. 2022; Available at: https://covid19.who.int/. Accessed 05 April 2022.
- 4. Jassat W, Abdool Karim SS, Mudara C, Welch R, Ozougwu L, Groome MJ, Govender N, von Gottberg A, Wolter N, Wolmarans M, Rousseau P, Blumberg L, Cohen C. Clinical severity of COVID-19 in patients admitted to hospital during the omicron wave in South Africa: a retrospective observational study. The Lancet 2020; 10: e961-e969.
- 5. Young M, Crook H, Scott J, Edison P. Covid-19: virology, variants, and vaccines. BMJ Med. 2022 Apr 1;1(1):e000040. doi: 10.1136/bmjmed-2021-000040. PMID: 36936563; PMCID: PMC9951271.
- 6. Franch-Pardo I, Napoletano BM, Rosete-Verges F, Billa L. Spatial analysis, and GIS in the study of COVID-19. A review. Science of The Total Environment 2020; 739.
- 7. Bhadra A, Mukherjee A, Sarkar K. Impact of population density on Covid-19 infected and mortality rate in India. Modeling Earth Systems and Environment 2021; 7: 623–629.
- 8. Carozzi F, Provenzano S, Roth S. Urban Density and COVID-19. IZA Institute of Labour Economics. 2020.
- 9. Zuo C, Meng Z, Zhu F, Zheng Y, Ling Y. Assessing Vaccination Prioritization Strategies for COVID-19 in South Africa Based on Age-Specific Compartment Model. Front Public Health 2022.
- 10. Studenmund AH. Using Econometrics A Practical Guide. 1124 Pearson New International Edition. Person Education Limited: Essex, England, 2014, 1125.
- 11. Huang H, Shi H, Zordan M, Lo SM, Tsou JY. Investigating the Spatiotemporal Relationship between the Built Environment and COVID-19 Transmission. ISPRS International Journal of Geo-Information. 2023 Sep 27;12(10):390.

Thank You!

