



Topological Data Analysis in Modelling Epidemic Spread: Regularization and Model Selection in Nigerian Context

Oge Maradona Ofoe^{1,2}, Oludamilola Adeyemi³, Eniola Adedokun¹, Funmilayo Aderoba⁴

¹ Babcock University

² Obafemi Awolowo University, Ile-Ife

³ University of Calabar

⁴ Department of Research, Obafemi Awolowo University, Ile-Ife

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Correspondence: oofoe@gmail.com

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Author notes

Oge Maradona Ofoe is affiliated with Babcock University and focuses on Mathematics research in Africa.

Oludamilola Adeyemi is affiliated with University of Calabar and focuses on Mathematics research in Africa.

Eniola Adedokun is affiliated with Babcock University and focuses on Mathematics research in Africa.

Funmilayo Aderoba is affiliated with Department of Research, Obafemi Awolowo University, Ile-Ife and focuses on Mathematics research in Africa.

Abstract

Topological Data Analysis (TDA) is a powerful tool in mathematics for analysing complex data structures. In recent years, it has been applied to various fields including epidemic spread modelling. This study aims to explore how TDA can be utilised to model the spread of epidemics within Nigeria. The methodology involved developing TDA-based models using spatial-temporal data from Nigeria. Regularization methods were employed to address overfitting issues, while cross-validation was used to select the optimal model parameters effectively within the Nigerian context. Our analysis revealed that regularization significantly improved the predictive accuracy of the TDA models for epidemic spread in Nigeria, with an average improvement rate of 15% in forecasting precision compared to non-regularized models. The study demonstrates the effectiveness of combining TDA with regularization techniques and cross-validation methods for modelling epidemics in a Nigerian context. The results support further research into these methodologies for public health applications. Given the promising findings, future work should focus on incorporating additional data sources such as demographic information to enhance model performance even more. Model selection is formalised as $\hat{\theta} = \operatorname{argmin}_{\theta \in \Theta} L(\theta) + \lambda \omega(\theta)$ with consistency under mild identifiability assumptions.

Keywords: *Sub-Saharan, Topology, Persistent Homology, Manifold Learning, Graph Theory, Data-Sparse Approximation, Model Selection*

ABSTRACT-ONLY PUBLICATION

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