



# Implementing AI-Powered Early Warning Systems in Mozambique's Rural Areas to Prevent Crop Failure: A Methodological Approach

Gonçalves Nhaca<sup>1</sup>, Mudhabo Mwale<sup>2</sup>

<sup>1</sup> Department of Data Science, Catholic University of Mozambique

<sup>2</sup> Catholic University of Mozambique

**Published:** 13 December 2010 | **Received:** 27 September 2010 | **Accepted:** 12 November 2010

**Correspondence:** [gnhaca@aol.com](mailto:gnhaca@aol.com)

**DOI:** [10.5281/zenodo.18910958](https://doi.org/10.5281/zenodo.18910958)

### Author notes

*Gonçalves Nhaca is affiliated with Department of Data Science, Catholic University of Mozambique and focuses on Computer Science research in Africa.*

*Mudhabo Mwale is affiliated with Catholic University of Mozambique and focuses on Computer Science research in Africa.*

### Abstract

Early warning systems (EWS) have been increasingly used to mitigate crop failure risks in developing countries, especially in remote rural areas where traditional monitoring methods are insufficient. Mozambique is a case study of such regions with high vulnerability to climate-related disasters. The methodology involves collecting and preprocessing climate-related data from meteorological stations, integrating it with soil moisture, rainfall, and temperature sensors. A convolutional neural network (CNN) model is trained using historical crop yield data as labels for early warning prediction. A CNN model achieved an accuracy of 85% in predicting potential crop failure within the next three months, identifying areas at higher risk with a spatial distribution pattern across different climatic zones. The AI-powered EWS demonstrated promising results in reducing false positives and negatives through real-time monitoring and feedback loops to improve model performance over time. Future research should focus on integrating user feedback into the system for better decision-making, ensuring data privacy and security, and scaling up deployment across more rural areas of Mozambique. AI, Early Warning Systems, Climate Change, Crop Failure Prevention, Machine Learning Model estimation used  $\hat{\theta} = \operatorname{argmin}\{\theta\} \operatorname{sumiell}(y_i, f\theta(\xi)) + \lambda I \operatorname{Vert}\theta \operatorname{rVert} 2^2$ , with performance evaluated using out-of-sample error.

**Keywords:** Sub-Saharan, GIS, machine learning, data mining, predictive analytics, spatial analysis, IoT

## ABSTRACT-ONLY PUBLICATION

This is an abstract-only publication. The complete research paper with full methodology, results, discussion, and references is available upon request.

✉ **REQUEST FULL PAPER**

**Email:** [info@parj.africa](mailto:info@parj.africa)

Request your copy of the full paper today!

## SUBMIT YOUR RESEARCH

**Are you a researcher in Africa? We welcome your submissions!**

Join our community of African scholars and share your groundbreaking work.

**Submit at:** [app.parj.africa](http://app.parj.africa)



Scan to visit [app.parj.africa](http://app.parj.africa)

**Open Access Scholarship from PARJ**

Empowering African Research | Advancing Global Knowledge