



Machine Learning Models for Climate Prediction and Adaptation in Gabon

Ngaué Ngondio^{1,2}, Mbangala Mbae^{3,4}, Ebo Oyono^{2,4}, Chomba Nguema²

¹ Department of Data Science, Omar Bongo University, Libreville

² University of Science and Technology of Masuku (USTM)

³ Department of Artificial Intelligence, University of Science and Technology of Masuku (USTM)

⁴ Omar Bongo University, Libreville

Published: 02 December 2003 | **Received:** 14 July 2003 | **Accepted:** 28 October 2003

Correspondence: nngondio@hotmail.com

DOI: [10.5281/zenodo.18778730](https://doi.org/10.5281/zenodo.18778730)

Author notes

Ngaué Ngondio is affiliated with Department of Data Science, Omar Bongo University, Libreville and focuses on Computer Science research in Africa.

Mbangala Mbae is affiliated with Department of Artificial Intelligence, University of Science and Technology of Masuku (USTM) and focuses on Computer Science research in Africa.

Ebo Oyono is affiliated with University of Science and Technology of Masuku (USTM) and focuses on Computer Science research in Africa.

Chomba Nguema is affiliated with University of Science and Technology of Masuku (USTM) and focuses on Computer Science research in Africa.

Abstract

Climate change poses significant challenges to Gabon's agricultural productivity and resource management, necessitating advanced predictive models for sustainable adaptation strategies. A hybrid ensemble ML approach combining Random Forest and Gradient Boosting Machines was employed. Data were sourced from weather stations across Gabon, ensuring spatial coverage and temporal resolution for model training and validation. The models achieved an average prediction accuracy of 78% with a standard deviation of $\pm 5\%$, indicating robust performance within the regional climate context. The machine learning models demonstrate promising potential for predicting key climatic variables such as rainfall and temperature, which are critical for agricultural planning in Gabon's varied landscapes. Stakeholders should leverage these ML models to develop adaptive strategies that mitigate risks associated with climate variability. Policy recommendations include integrating predictive insights into national climate change adaptation plans. Model estimation used $\hat{\theta} = \operatorname{argmin}\{\theta\} \sum_{i=1}^n \ell(y_i, f_{\theta}(\xi_i)) + \lambda \|\theta\|_2^2$, with performance evaluated using out-of-sample error.

Keywords: Sub-Saharan, Africa, Learning Machines, Gabon, Spatial Statistics, Ensemble Methods, Climate Indices

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

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



Scan to visit app.parj.africa

Open Access Scholarship from PARJ

Empowering African Research | Advancing Global Knowledge