Application of Convolutional Neural Networks for Aeromagnetic Anomaly Classification in the Kenyan Greenstone Belt for Gold Exploration

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Background

1. Introduction

•The use of Machine learning (ML), particularly deep learning models like Convolutional Neural Networks (CNNs), offers an efficient and scalable alternative for anomaly detection and classification in geospatial data.

•The use of CNN in aeromagnetic data analysis and interpretation can help so much in exploration of

3. Results





greenfield prospect that may not be delineated when traditional manual interpretation methods are applied.

2. Problem statement

•Manual interpretation of aeromagnetic anomalies is time consuming and prone to inconsistencies.

There is a lack of automated systems tailored for aeromagnetic data analysis in Kenyan contexts.
Identifying subtle geophysical signatures indicative of gold and base metals mineralization remains challenging

Current Work

1. Objectives

To pre-process and convert raw aeromagnetic survey data into interpretable grids.
To apply and evaluate CNN-based models for classifying magnetic anomalies associated with





Fig 5. Manual interpretation output

Conclusion & Expectations

•CNNs are viable tools for automated classification of aeromagnetic anomalies in mineral exploration settings.

•This approach provides a faster and more consistent method for highlighting greenfield prospective zones in under-explored regions and remote areas.

•Future work includes: Expanding the dataset to include other geological formation in Kenya and geophysical parameters (e.g., gravity).

mineralized structures.

•To develop a reproducible ML pipeline that supports future mineral exploration efforts in similar greenstone terrains.

2. Methodology

Data Acquisition: Aeromagnetic, Geochemical, Maps
↓
Data Preprocessing: RTP, Filters, Normalization
↓
CNN Model Design: VGG/ResNet Architecture
↓
CNN Training: Cloud Platform
↓
Model Validation: ROC, Accuracy, Confusion Matrix
↓
Multi-Source Integration: Geochemical Data

Fig. 3 Analytic Signal

Fig 4. Possible CNN output map



References

LeCun, Y. et al. (1998). Gradient-Based Learning Applied to Document Recognition.
Proceedings of the IEEE
Ministry of Mining, Kenya. Aeromagnetic
Survey Data Repository.
Smith, J. et al. (2020). Deep Learning
Applications in Geophysics. Journal of
Applied Geophysics.

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Generate Probability Maps: Mineral Prospectivity

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Interpretability Analysis: XAI (Grad-CAM)

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Collaboration & Dissemination: Share Results, Publish

Fig. 1 workflow



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