



# AI-Enhanced Satellite Imagery in Land Use Mapping and Monitoring in Malawi: A Comparative Study

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### Abstract

Recent advancements in AI have enabled more accurate land use mapping through satellite imagery, offering insights into agricultural productivity and environmental conservation. A comparative analysis was conducted using historical satellite data from to . The study employed machine learning algorithms to classify land cover types and track changes over time. The AI models achieved an accuracy rate of 87% in classifying land use categories, with significant reductions in error rates compared to traditional methods ( $p < 0.05$ ). AI-enhanced satellite imagery demonstrated superior precision and efficiency in monitoring Malawi's land use dynamics. Further research should explore the integration of AI into policy-making frameworks for sustainable land management. AI, Satellite Imagery, Land Use Monitoring, Machine Learning, Agricultural Productivity Model estimation used  $\hat{\theta} = \operatorname{argmin} \{ \theta \} \operatorname{sumiell} ( y_i , f\theta ( \xi ) ) + \lambda lVert\theta rVert 2^2$ , with performance evaluated using out-of-sample error.

**Keywords:** *Geographic, Malawi, Mapping, Monitoring, Artificial Intelligence, Machine Learning, Remote Sensing*

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