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- [Published: 06 May 2020](#)

Assessment of drivers of forest changes using multi-temporal analysis and boosted regression trees model: a case study of Nyeri County, Central Region of Kenya

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[Modeling Earth Systems and Environment](#) volume 6, pages1657–1670(2020)[Cite this article](#)

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Abstract

The Central Region of Kenya has undergone significant changes in land cover due to a broad range of drivers. These changes are more pronounced in forestland conversions. Past researches within the study area have identified drivers of land cover change without quantifying the influence of these drivers. Predictor variables include population density, precipitation, elevation, slope, forest fires, soil texture, proximity to roads, rivers and towns. Land cover changes were analyzed using multi-temporal land cover maps between year 1990 and 2014. Boosted regression trees model was applied to determine the significant drivers and quantify their relative influence on key forestland transitions. The local and spatial influence of the drivers has further been analyzed by geographical weighted regression using coefficients determined at each sample point. Significant land cover changes continuously occurred over the study period. Forestland reduced from 38.90% in 1990 to 38.14% in 2014. Grassland reduced from 32.59 to 22.57%, cropland increased from 28.05 to 38.83% and wetland changed

from 0.07 to 0.04%. Other land which constitutes of bare land and built up increased from 0.38 to 0.42%. The results show population density had the highest contribution to forestland changes throughout the study period, with a minimum contribution of 20.02% to a maximum of 26.04%. Other significant variables over the study period are precipitation, slope, elevation and the proximity variables. The results indicate that the relative influence of the drivers to forestland conversion varies with time, location and type of transition.

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Acknowledgements

We acknowledge the Ministry of Environment and Forestry, Kenya for providing Landsat images and land cover maps. We are grateful to two anonymous reviewers for valuable comments on the initial draft of the manuscript.

Funding

This study was part of System for Land-based Emissions Estimation in Kenya (SLEEK) Project, funded by the Australian Government in partnership with the Government of Kenya.

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Contributions

NM contributed to formal analysis, methodology and writing of first draft. HW supervised, reviewed and edited the manuscript. CM and JK contributed to conceptualization of project and supervision of research. FN

contributed to project administration and acquisition of funding. All authors commented on the manuscript.

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Ethics declarations

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All authors declare that they have no conflict of interest.

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- Received: 30 December 2019
- Accepted: 10 April 2020
- Published: 06 May 2020
- Issue Date: September 2020
- DOI: <https://doi.org/10.1007/s40808-020-00781-2>

Keywords

- Change analysis
- Drivers of change
- Relative influence
- Spatial influence
- Central Kenya

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