

Using Topographic Attributes to Predict the Understorey Structure of a Wet Eucalypt Forest

Bechu Kumar Vinwar Yadav (✉ bechu.yadav@utas.edu.au)

University of Tasmania, School of Technology, Environments and Design, 4 Discipline of Geography and Spatial Sciences, Private Bag 70, Hobart, Tasmania 5 7001, Australia. <https://orcid.org/0000-0002-1076-6099>

Arko Lucieer

University of Tasmania, School of Technology, Environments and Design, Discipline of Geography and Spatial Sciences

Gregory J. Jordan

University of Tasmania, School of Natural Sciences, Biological Sciences

Susan C. Baker

University of Tasmania, School of Natural Sciences, Biological Sciences

Research

Keywords: Airborne LiDAR, Digital terrain model, Topographic attributes, 44 Geology, Understorey structure, Random forest, Variable importance

Posted Date: December 29th, 2020

DOI: <https://doi.org/10.21203/rs.3.rs-135008/v1>

License:  This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

Abstract

Background: Forest understorey structure is an important component of forest ecosystems that affects forest-dwelling species, nutrient cycling, fire behaviour, biodiversity, and regeneration capacity. Mapping the structure of forest understorey vegetation with field surveys or high-resolution LiDAR data is costly. We tested whether landscape topography and underlying geology could predict the understorey structure of a 19 km² area of wet eucalypt primary forest located at the Warra Long Term Ecological Research Supersite, Tasmania, Australia. In this study, we used random forest regressions based on twelve topographic attributes derived from digital terrain models (DTMs) at various resolutions and a geology variable to predict the densities of three understorey layers compared to density estimates from a high resolution (28.66 points/m²) LiDAR survey.

Results: We predicted the vegetation density of three canopy strata with a high degree of accuracy (validation root mean square error ranged from 8.97% to 13.69%). 30 m resolution DTMs provided greater predictive accuracy than DTMs with higher spatial resolution. Variable importance depended on spatial resolutions and canopy strata layers, but among the predictor variables, geology generally produced the highest predictive importance followed by solar radiation. Topographic position index, aspect, and SAGA wetness index had moderate importance.

Conclusions: This study demonstrates that geological and topographic attributes can provide useful predictions of understorey vegetation structure in a primary forest. Given the good performance of 30 m resolution, the predictive power of the models could be tested on a larger geographical area using lower density LiDAR point clouds. This study should help in assessing fuel loads, carbon stores, biomass, and biological diversity, and could be useful for foresters and ecologists contributing to the planning of sustainable forest management and biodiversity conservation.

Full Text

This preprint is available for [download as a PDF](#).

Figures

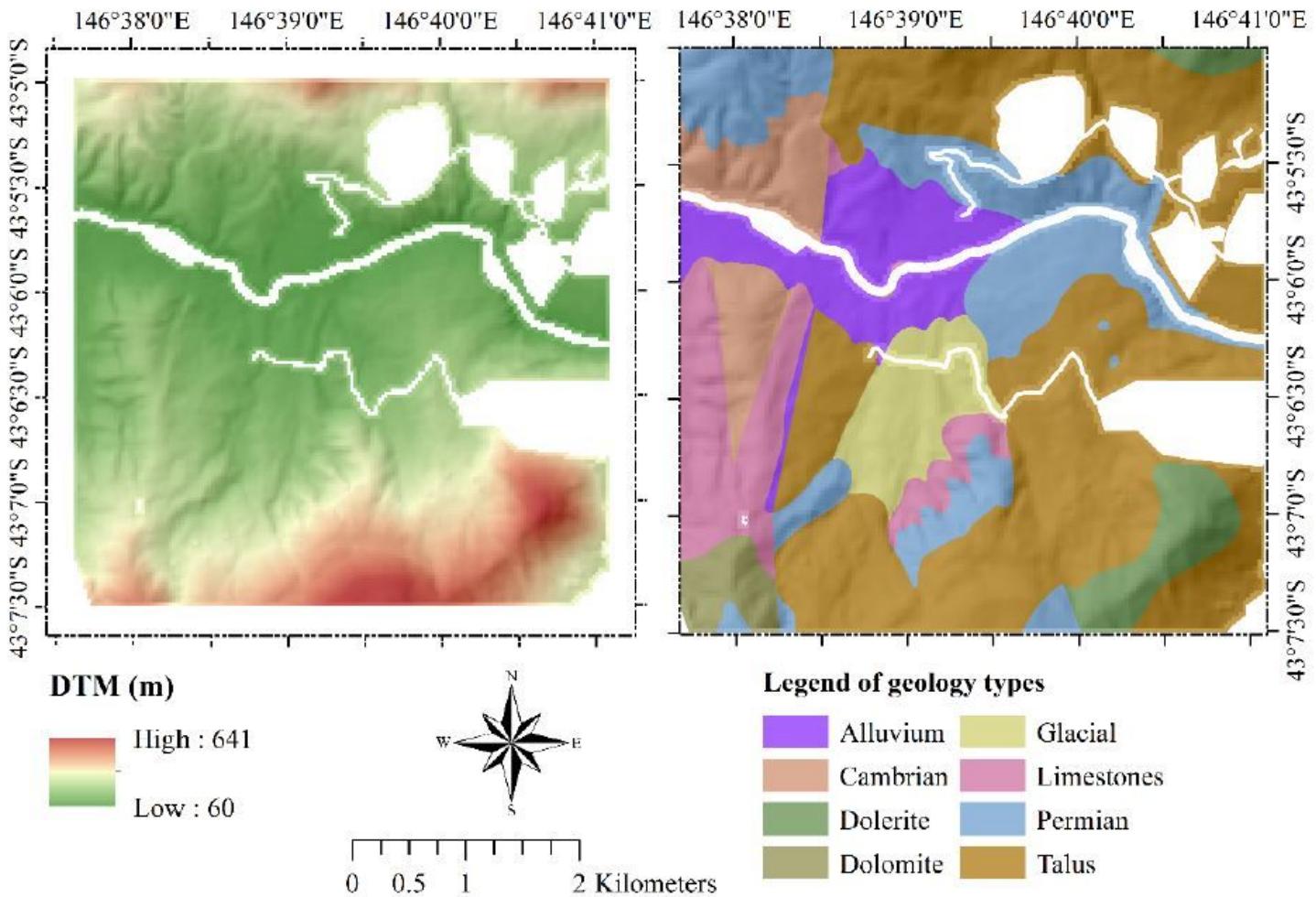


Figure 1

Maps of the study site showing the 30 m resolution digital terrain model (left) and geology (right). White areas removed from the analysis were roads, rivers, and previously harvested sites. Abbreviations of geology types are Quaternary alluvium (Alluvium), Cambrian siliceous sediments (Cambrian), Jurassic dolerite (Dolerite), Neoproterozoic dolomite (Dolomite), Glacial tills (Glacial), Permian sediments (Permian), and dolerite talus (Talus).

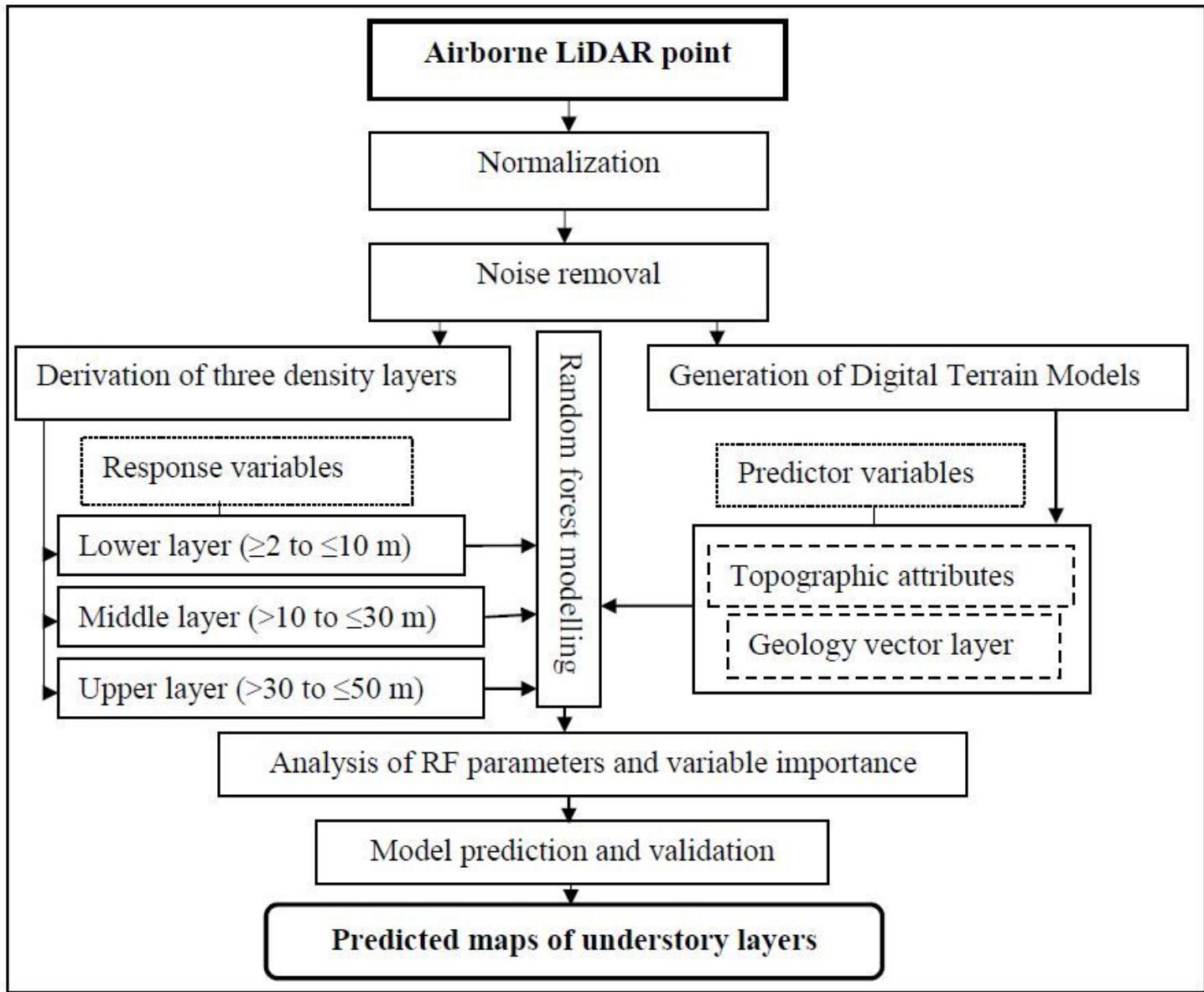


Figure 2

Workflow for testing the capacity for topography and geology to predict the density of vegetation in three canopy layers of a mature forest landscape.

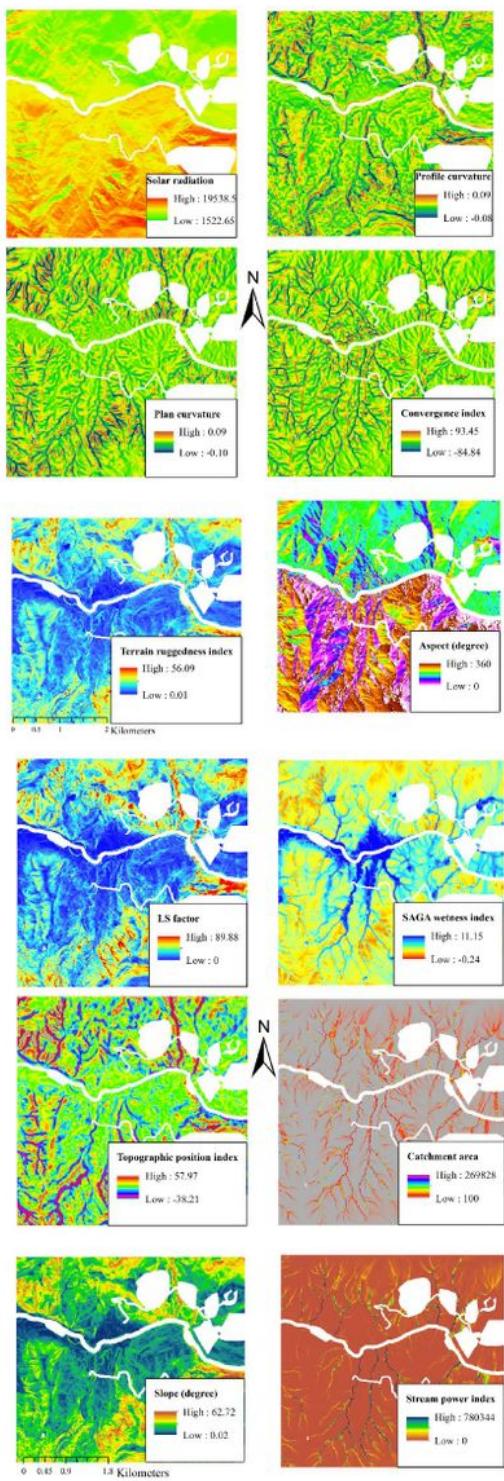


Figure 3

Maps of the twelve topographic attributes extracted from DTMs.

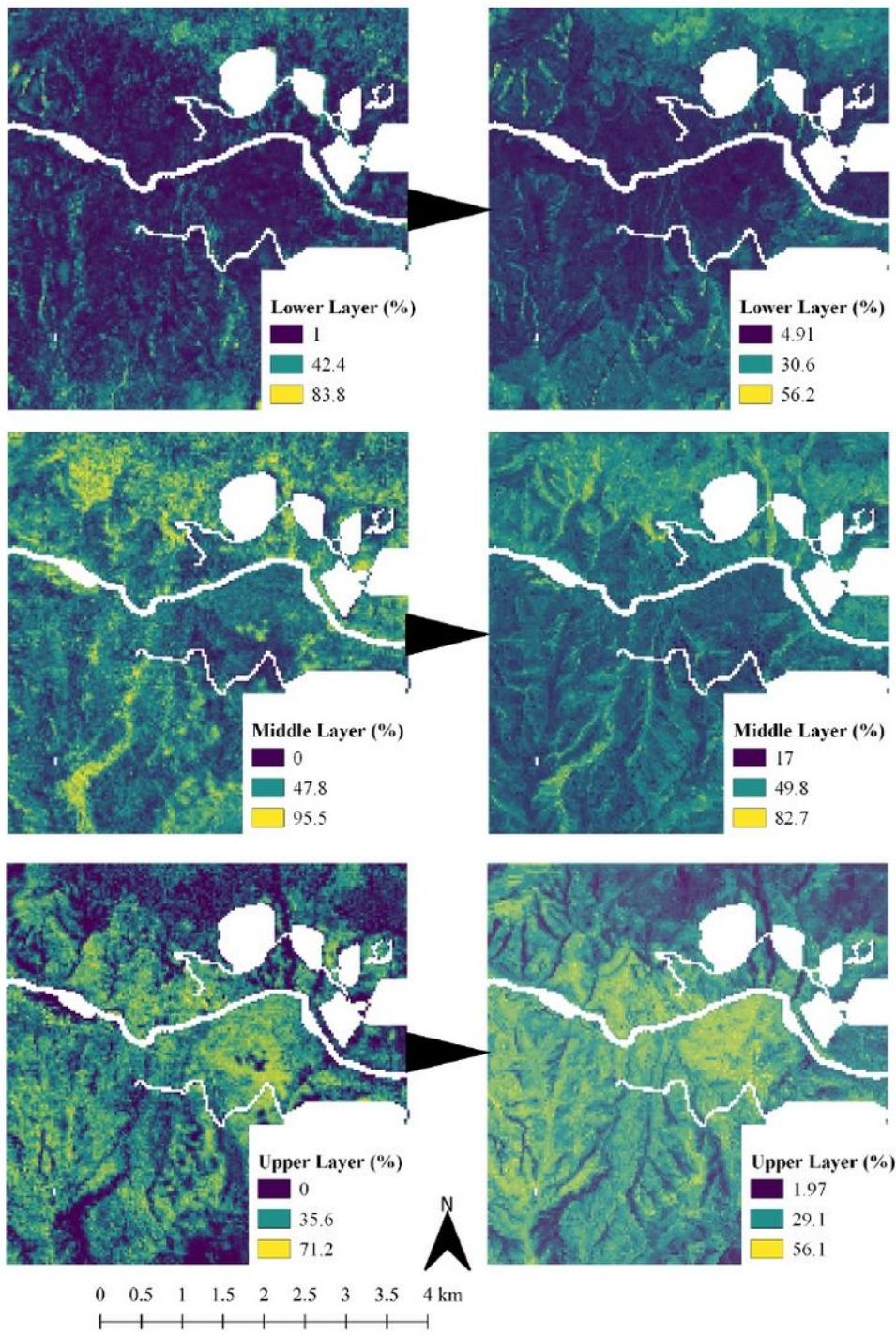


Figure 4

Maps comparing the density of forest understorey layers that were observed with high-resolution LiDAR data (left) with those predicted based on topography and geology (right) using 30 m resolution. Roads, rivers, and harvested areas appear as white.

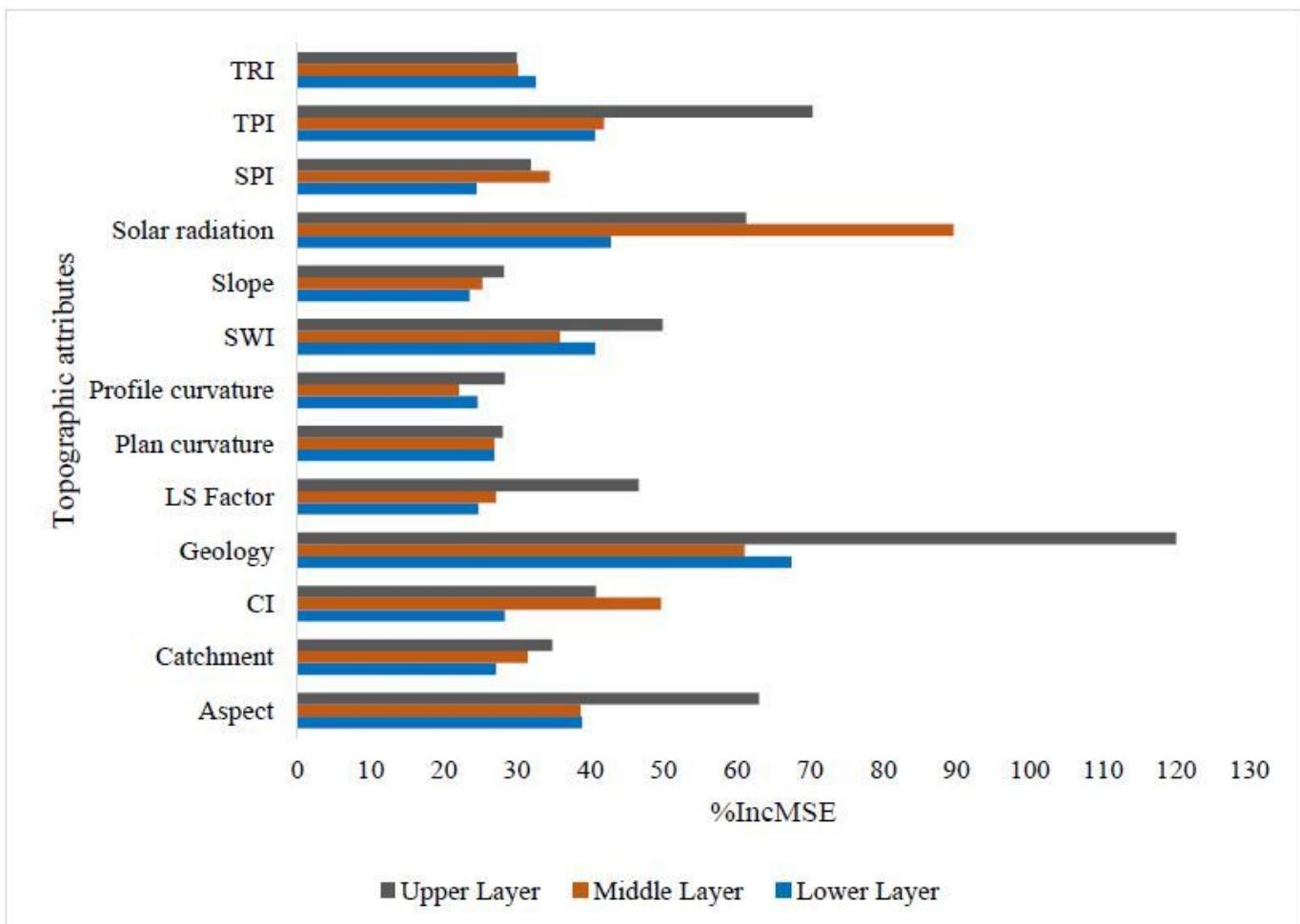


Figure 5

Variable importance scores (percentage increase in mean square error, %IncMSE if the variable is eliminated from models) for topographic and geological attributes used in predicting the density of three vegetation canopy layers at 30 m spatial resolution. (Acronyms: Terrain Ruggedness Index (TRI), Topographic Position Index (TPI), Stream Power Index (SPI), SAGA Wetness Index (SWI), and Convergence index (CI)).

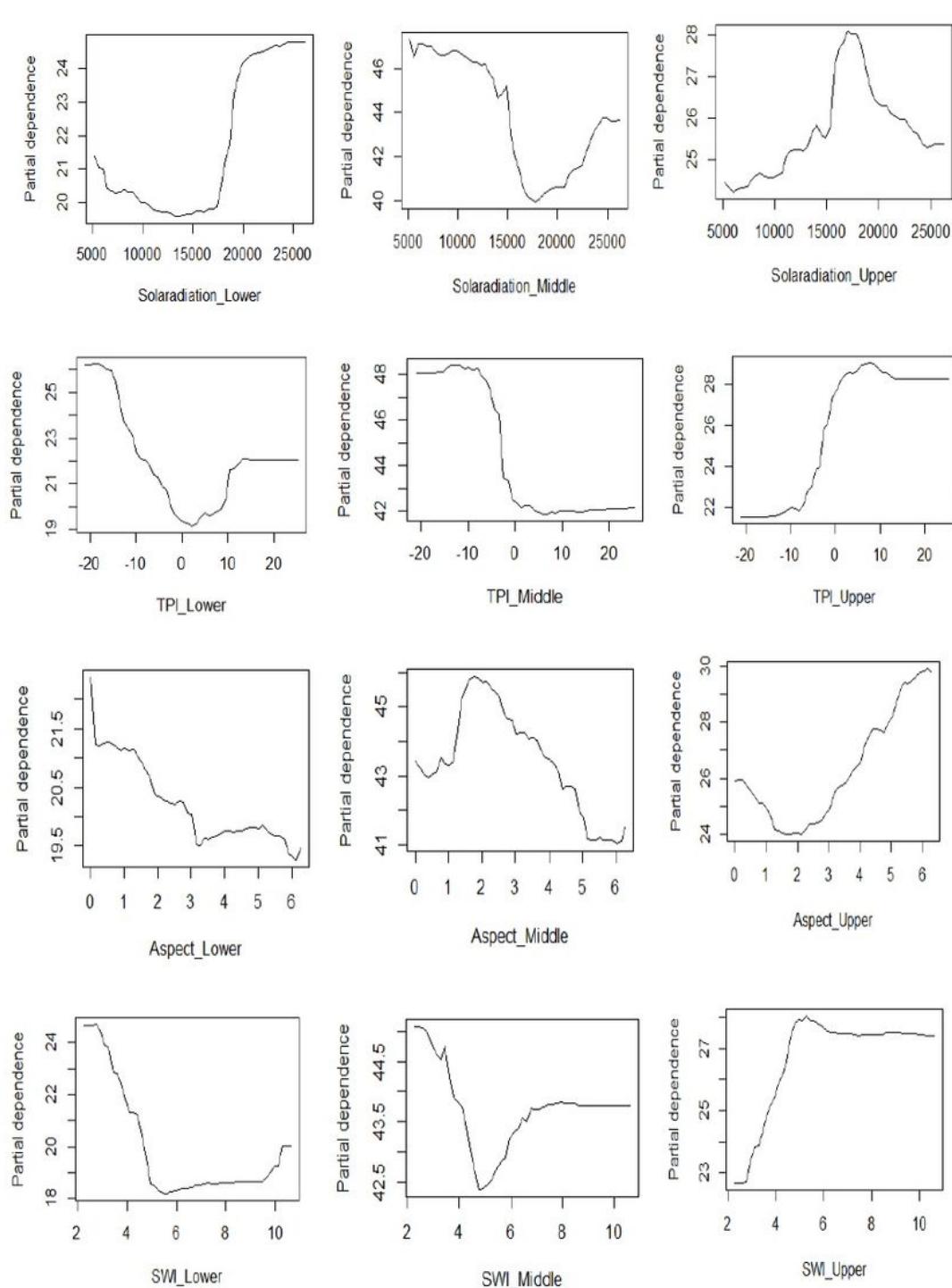


Figure 6

Partial dependence plots of solar radiation, terrain position index (TPI), aspect, and SAGA wetness index (SWI) for random forest models of the three forest layers. Higher partial dependence values indicate a greater influence of the variables on the model.

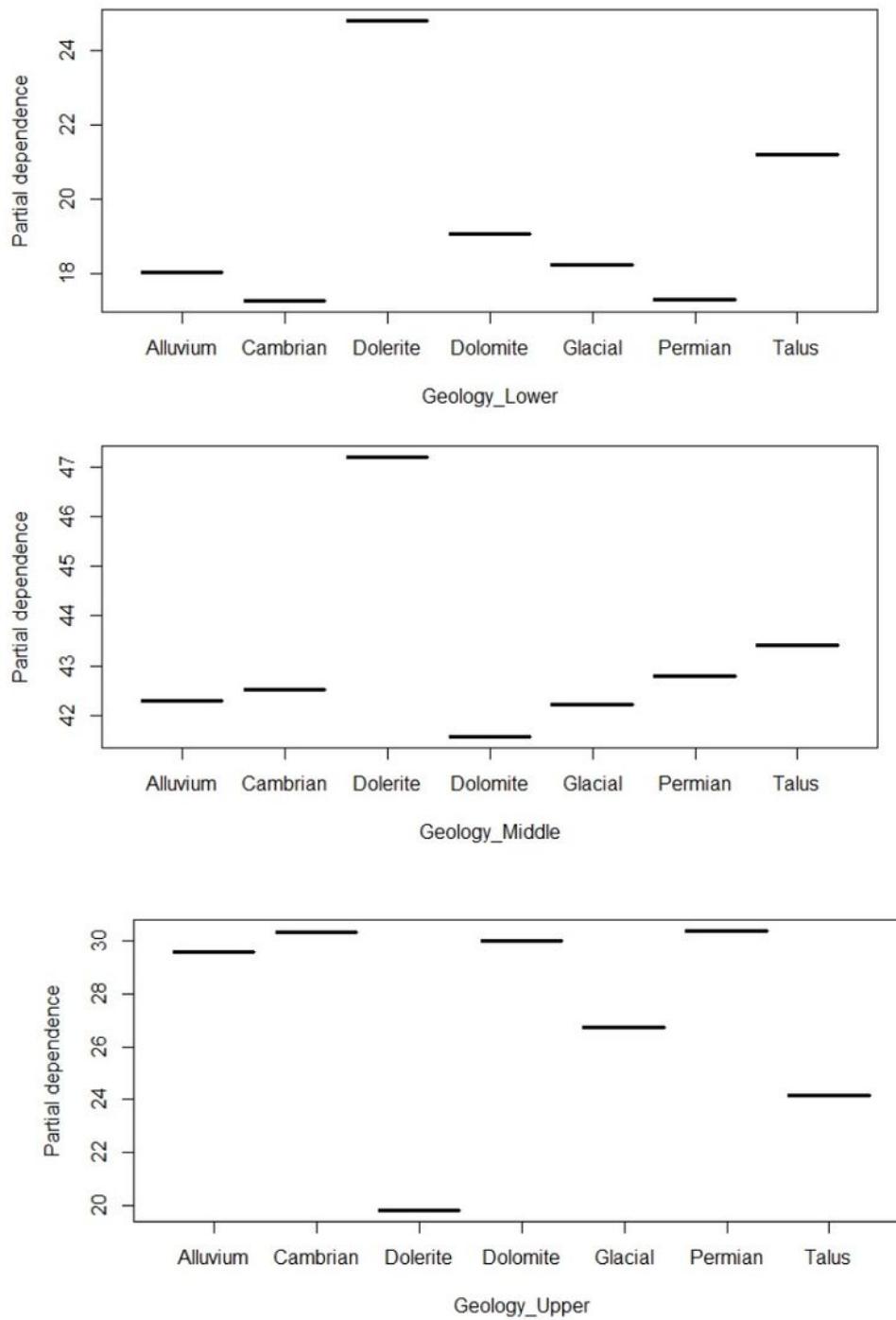


Figure 7

Partial dependence plots of geology. The acronyms used are Quaternary alluvium (Alluvium), Cambrian siliceous sediments (Cambrian), Jurassic dolerite (Dolerite), Neoproterozoic dolomite (Dolomite), Glacial tills (Glacial), Permian sediments (Permian), and Dolerite talus (Talus). All Permian sedimentary rocks were integrated into a single category; the carbonaceous rock types Ordovician limestones and Cambrian

dolomite were combined; Dolerite boulders were merged with Dolerite talus; Quaternary sediments were merged with Quaternary alluvium.