

Telkom University due to resource constraints, limiting the spatial coverage of the dataset. Additionally, the overall dataset size remains relatively small, which may impact the generalizability of the model to other regions with different environmental conditions. Future work should aim to expand the dataset to include more diverse imagery from various regions, improving model robustness and performance across different landscapes.

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