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Wildfire Detection Using Machine Learning

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Abstract: *Due to climate change globally Forest fire disasters are recently getting lots of attention. Because climate changes are rapidly changing the fire patterns on Earth. Prediction of fire activities in the forest guides the authorities to make optimal, efficient, and sound decisions in fire management. This paper aims to summarize recent trends in the forest fire events prediction, detection, spread rate, and mapping of the burned areas. Furthermore, fire emissions in terms of smoke also put the Earth's public health and ecological system at greater risk. Hence, future policy making can be more accurate in saving billions of dollars, improving the healthy environment and ecological cycle for the inhabitants of this Earth. This paper provides a comprehensive review of the usage of different machine learning algorithms in forest fire or wildfire management.*

Keywords: Machine Learning, Wildfire, Forest Fires, Fire Susceptibility, Forecasting

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