
EVALUATING WILD FIRE DETECTION

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ABSTRACT

We are special. The fire detection system used in this study is based on CNNs (Convolutional Neural Networks). With conventional methods of smoke detectors integrated into structures, fire detection can be challenging, expensive and awkward owing to outmoded design and technology. The application of artificial intelligence for video identification and alerting based on CCTV footage is critically examined in this paper. This experiment makes use of a collection of fire-filled, self-generated video frames. Before creating machine learning models using CNNs, data is pre-processed. Use test data sets as a starting point to validate theories and record experiments. The objective of this project is to create a low-cost, highly accurate device that may be utilized in virtually any fire detection situation. This study also suggests a system and method for deploying a wireless monitoring network to detect forest fires early.

Keywords: Fire Alarm, Convolutional Neural Networks, ML, Security Cameras, Object Detection.

I. INTRODUCTION

Along with the loss of ecosystems brought on by massive emissions of smoke and carbon dioxide (CO₂), there have also been significant, long-lasting harms to the environment and climate (climate change is 30% carbon dioxide). Forest fires are responsible for this ambiance. Wildfires can cause long-term harm, including as conflicts with local weather patterns, global warming, and the extinction of endemic plant and animal species. Forest fires are especially deadly because they frequently happen in remote, abandoned, or poorly maintained areas that are heavily forested with trees, dry, parched lumber, leaves, and other fuels.

These components combine to create a highly flammable substance that serves as fuel for the spread of the fire and the ideal environment for its initial ignition. The fire may start owing to natural or human reasons, such as smoking or barbecue. Perhaps it's the heat of a summer's day, or perhaps the broken glass acts as a gigantic lens, concentrating the sun's rays on one area for a considerable amount of time before igniting.

II. METHODOLOGY

The study's suggested approaches are divided into numerous stages. Following are the steps: Getting the dataset, preprocessing the data, and processing the data are the first three steps. Model construction, feature extraction, and validation and testing make up the remaining three steps.

Dataset Acquisition: However, for training and testing, specially created films will be utilized for convenience of use. The data is video frames acquired from CCTV footage. These fire-related videos' compilation was a challenging task. Then, the fire-free and fire-prone frames are kept in different storage locations. Next, training and testing modes are separated from the data set. However, be cautious as incorrect data might ruin the neural network's output and prevent it from creating an accurate system.

Data Preparation: The following step in creating an excellent ML model is data preprocessing. Here, information is either cleansed and processed or is ready for usage. Preprocessing data entails removing unwanted objects and background noise from the frame. The algorithm must request relevant data or it may produce unfavorable results.

III. MODELING AND ANALYSIS

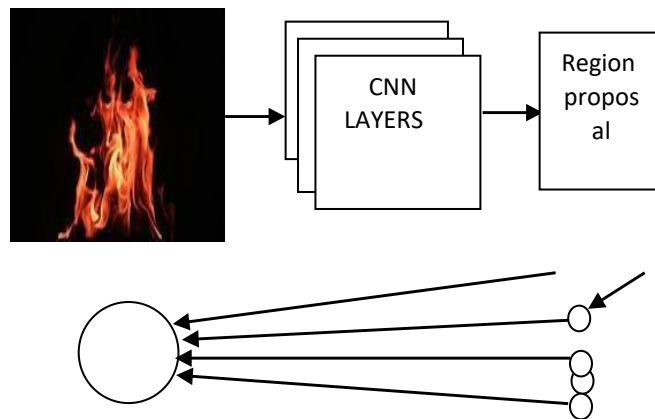


Figure 1: Architecture of CNN

IV. FUTURE ENHANCEMENT

Identifying obstacles is essential for improving wildfire detection methods. Issues including data integration, interagency cooperation, public involvement, and the requirement for defined processes are all mentioned in this section. The report also highlights future research directions, including as combining different detection techniques, strengthening early warning systems, and post-fire damage assessment.

V. CONCLUSION

A wireless sensor network called the Forest Fire Alarm System (FAS) employs machine learning to find forest fires. The analysis is carried out at both sensor nodes and base stations to provide the most accurate data with the shortest delay, and it has been demonstrated to be a practical method for detecting fires in forests. The system can adjust to any weather, climate, or location thanks to the threshold ratio introduced by the sensor node. Even if you don't have access to the internet, you can set up the transceiver anywhere in the woods because it is connected to the existing network. It is simple to set up as a stand-alone system for a very long time using rechargeable batteries and a second solar power source powering the main power source. The system, which was connected to the communication network, delivered notifications to the appropriate persons with a lot shorter delay during a few tests in actual tropical woods than the present techniques.

VI. REFERENCES

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