



ISSN: 2321-2152

IJMECE

*International Journal of modern
electronics and communication engineering*

E-Mail

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editor@ijmece.com

www.ijmece.com

Enhanced Forest Fire Detection Using Deep Learning

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ABSTRACT_ Utilising image processing technology to identify fires and notify people is the project's ultimate goal. Many automated fire alarm systems, such as the rather restrictive sensor method, are now operational. A camera plus the PyCharm IDE make up the hardware required to execute the project. Python and the open-source CV library are the backbone of the programming used for image processing. A camera acts as an input device, recording footage from the surrounding area and feeding it into the system for processing. The main focusses of the project are its colour model, computer vision, machine learning, and image processing tools, as well as its functioning fire detection algorithm. We want to fix the problems with the existing systems and provide a reliable system that can detect fires quickly, work in many environments, and save a lot of lives and resources.

1.INTRODUCTION

For the most part, fire mishaps cause monetary and biological harm as well as jeopardizing individuals' lives. To stay away from the fire's catastrophes, many early fire-recognition strategies have been investigated and a large portion of them depend on molecule examining, temperature inspecting, relative mugginess testing, air straightforwardness testing, smoke examination, notwithstanding the customary bright and infrared fire identifiers. Be that as it may, those identifiers either should be set nearby a fire or can't give the extra data about the method involved with consuming, like fire

area, size, developing rate, etc. In this manner, they are not generally solid since energy discharge of non-fires. The paper presents an early alarm raising strategy in light of video handling. The essential thought of the proposed of fire-recognition is to embrace another methodology for fire discovery which depends on some PC vision strategies. In light of the issues, we imagine that visual sensors, for example, video information obtained from CCTV or computerized camera turned into an elective way in leading fire location sensor, which was assessed that by utilizing video information, the discovery results can be all the more rapidly,

precisely, and really from the side inclusion region and can be applied to screen the climate indoor and out-entryway. a RGB (red, green, blue) model based chromatic and issue estimation for removing fire-pixels. The choice capability of fire-pixel is predominantly concluded by the power and immersion of R part. The separated fire-pixels will be checked in the event that it is a genuine fire by the two elements of development and turmoil

2.LITERATURE SURVEY

[1] Fire detection: The state-of-the-art NBS Technical Note, US.

Year of publication: 2021

Description: The document provides an overview of the principles of fire detection, types of detectors, and their operational characteristics. It highlights the importance of early detection in preventing fire-related fatalities and property damage. The technical note discusses different types of fire detection methods, including smoke detection, heat detection, and flame detection. It outlines the advantages and disadvantages of each method and emphasizes the need for reliable and accurate fire detection systems.

The document also covers the latest technologies in fire detection, including multi-sensor detectors, which can combine different sensing elements to improve the accuracy and reliability of fire detection.

Mainly:

The technical note emphasizes the importance of proper installation, maintenance, and testing of fire detection systems to ensure their effectiveness. It provides guidelines for the placement and installation of fire detectors in various types of buildings and environments.

In conclusion, the NBS Technical Note on **"Fire Detection: The State of the Art"** provides a valuable resource for fire safety professionals, engineers, and researcher

[2] Image processing-based forest fire detection, International Journal of Emerging Technology and Advanced Engineering.

Year of publication: 2022

Description: A clever methodology for woodland fire discovery utilizing the picture handling strategy is proposed.

A standard based variety model for fire pixel grouping is utilized. The proposed calculation utilizes RGB and YC b C r variety space. The ability of YC b C r color space to effectively distinguish luminance from chrominance makes it superior to RGB color space. On two sets of images, one of which contains fire, the proposed algorithm is tested for its effectiveness; The other has areas that look like fire. The algorithm's performance is calculated using standard techniques. The proposed technique has both a higher location rate and a lower deception rate.

Since the calculation is modest in calculation, it very well may be utilized for constant woodland fire discovery. Watchwords Backwoods fire discovery, picture handling, rule-based variety model, picture division.

[3] A probabilistic approach for vision-based fire detection in videos

Year of publication: 2020

Description: Computer vision is currently conducting research on automated fire detection. A novel approach to identifying fire in videos is the subject of our investigation in this paper. PC vision-based fire location calculations are normally applied in shut circuit TV observation situations with controlled foundation. In contrast, the proposed method can be used for both surveillance and automatic video classification in order to retrieve fire disasters from newscast content databases. The changes that occur from frame to frame in particular low-level features that describe potential fire regions are the focus of the proposed method. Color, area size, surface roughness, boundary roughness, and skewness within estimated fire regions are these characteristics.

In view of flashing and arbitrary qualities of fire, these highlights are strong discriminants. For robust fire recognition, the behavioral change of each of these features is evaluated, and the results are combined using a Bayes classifier. Additionally, the classification results are significantly enhanced by using prior knowledge of video fire events. For altered report recordings, the fire locale is typically situated in the focal point of the casings. The probability of fire as a function of position is modeled using this fact. The method's applicability was demonstrated through experiments.

3.PROPOSED SYSTEM

An image-based fire detection system based on computer vision-based methods was proposed in this system. We have gathered several consecutive frames from the original video, including both photos with and without fire. There are three primary phases to the suggested method: - moving pixel detection, evaluating the shape of fire-colored pixels in frames to find fire pixel in image; - fire pixel detection utilising RGB and YCbCr colour model. After applying the suggested strategy to film sequences, fire is found.

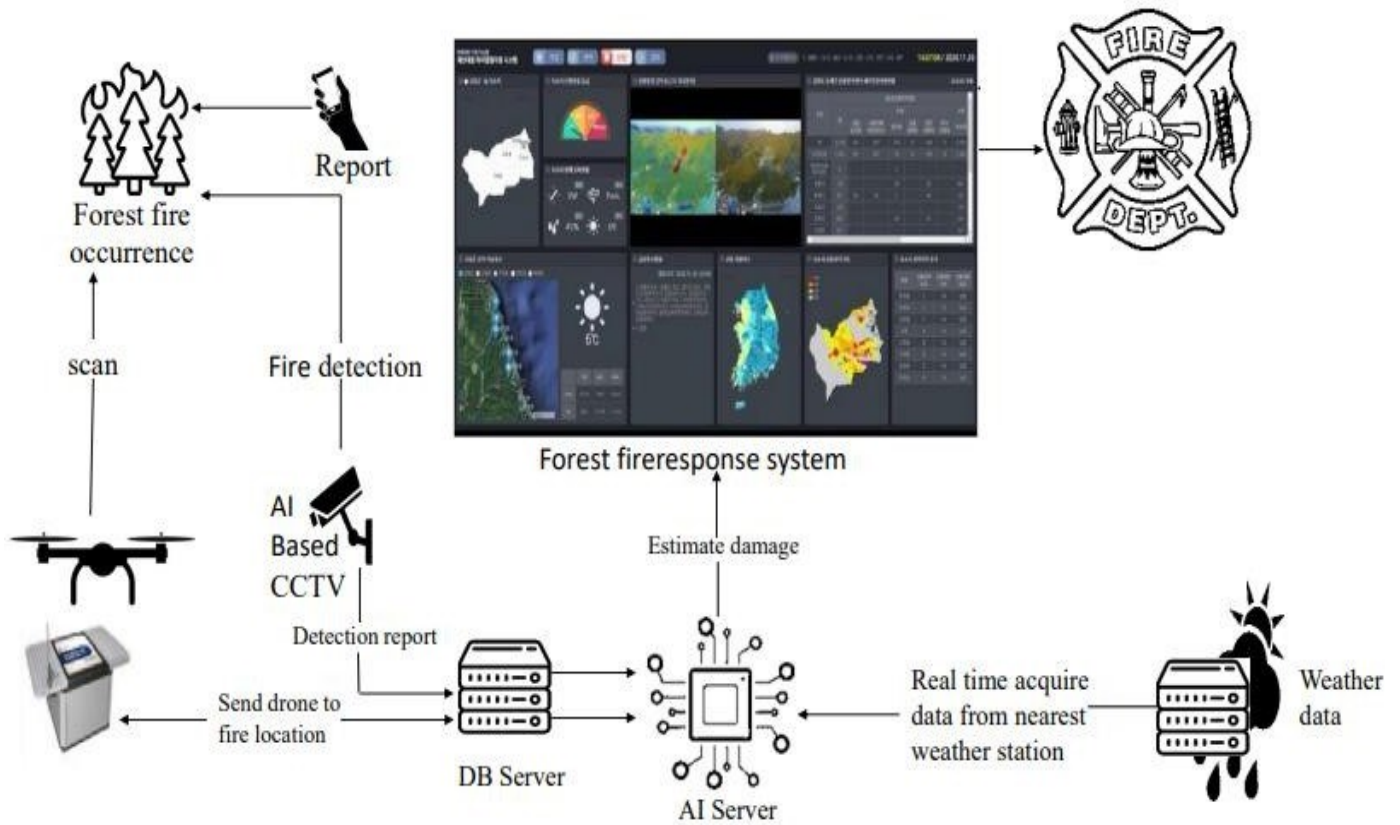


Figure 1: block diagram of deep learning- based forest fire response system

Figure 1 shows a block diagram of a deep learning-based forest fire response system, which includes a data acquisition module, a deep learning model for fire detection, a decision-making module, and a response module for initiating appropriate actions in response to detected fires.

3.1 IMPLEMENTATION

3.1.1 Image Acquisition

Image acquisition can be defined as the act of procuring an image from sources. This

can be done by hardware system such as cameras and datasets and also some encoders sensors also take place in this process.

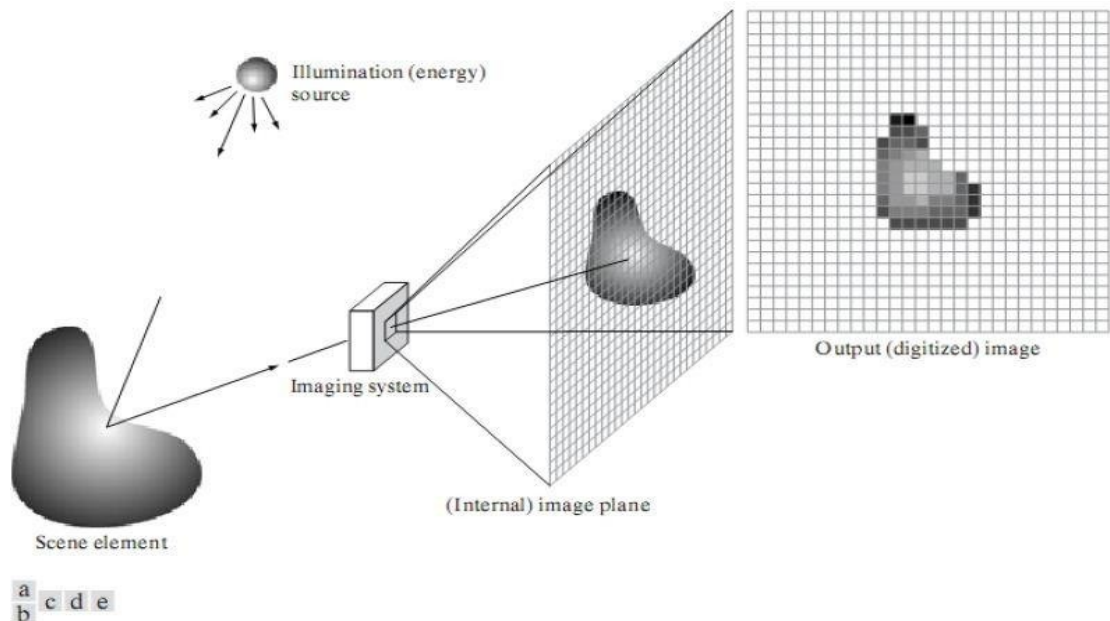


Figure 2: image acquisition

3.1.2 Pre Processing

In this step, the acquired images are pre-processed to enhance image quality and correct for any distortions or noise. This may involve techniques such as image filtering, color correction, and image registration.

Image processing

Image processing is a method to convert an image into digital form and perform some operations on it, in order to get an enhanced image or to extract some useful information from it. It is a type of signal dispensation in which input is image, like video frame or photograph and output may be image or characteristics associated with that image. Usually, Image Processing system includes treating images as two-dimensional signals while applying

already set signal processing methods to them.

It is among rapidly growing technologies today, with its applications in various aspects of a business. Image Processing forms core research area within engineering and computer science disciplines too.

Image processing basically includes the following three steps.

- Importing the image with optical scanner or by digital photography.
- Analyzing and manipulating the image which includes data compression and image enhancement and spotting patterns that are not to human eyes like satellite photographs.
- Output is the last stage in which result can be altered image or report

that is based on image analysis.

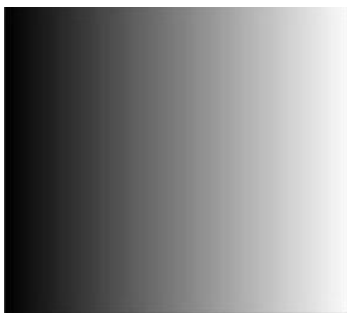
Digital Processing techniques help in manipulation of the digital images by using computers. As raw data from imaging sensors from satellite platform contains deficiencies.

3.1.3 Finding Image Details

After loading the image with the `imread()` function, we can then retrieve some simple properties about it, like the number of pixels and dimensions:

3.1.4 Image Thresholding

The concept of thresholding is quite simple. As discussed above in the image representation, pixel values can be any value between 0 to 255. Let's say we wish to convert an image into a binary image i.e., assign a pixel either a value of 0 or 1. To do this, we can perform thresholding. For instance, if the Threshold (T) value is 125, then all pixels with values greater than 125 would be assigned a value of 1, and all pixels with values lesser than or equal to that would be assigned a value of



0. Let's do that through code to get a better understanding.

Figure :5. Image used for Thresholding

3.1.5 Image Segmentation

Image segmentation is the process of dividing an image into multiple regions based on similarities in color, texture, or other features. This step is used to identify regions of interest in the image, such as objects or areas with specific characteristics.

3.1.6 Feature Extraction

In this step, key features of the segmented regions are extracted to represent the image in a more compact and meaningful way. These features may include shape, texture, color, or other properties that can be used to identify or classify the image.

3.1.7 Classification

The extracted features are used to classify the image into one or more predefined categories, based on the intended application of the image processing. This may involve using machine learning algorithms, such as neural networks or support vector machines, to train a model to recognize and classify images.

3.1.8 Post Pre-processing

Finally, the classified images may undergo post-processing, which may involve further enhancement, filtering, or compression to prepare them for storage or display.

4.RESULTS AND DISCUSSION

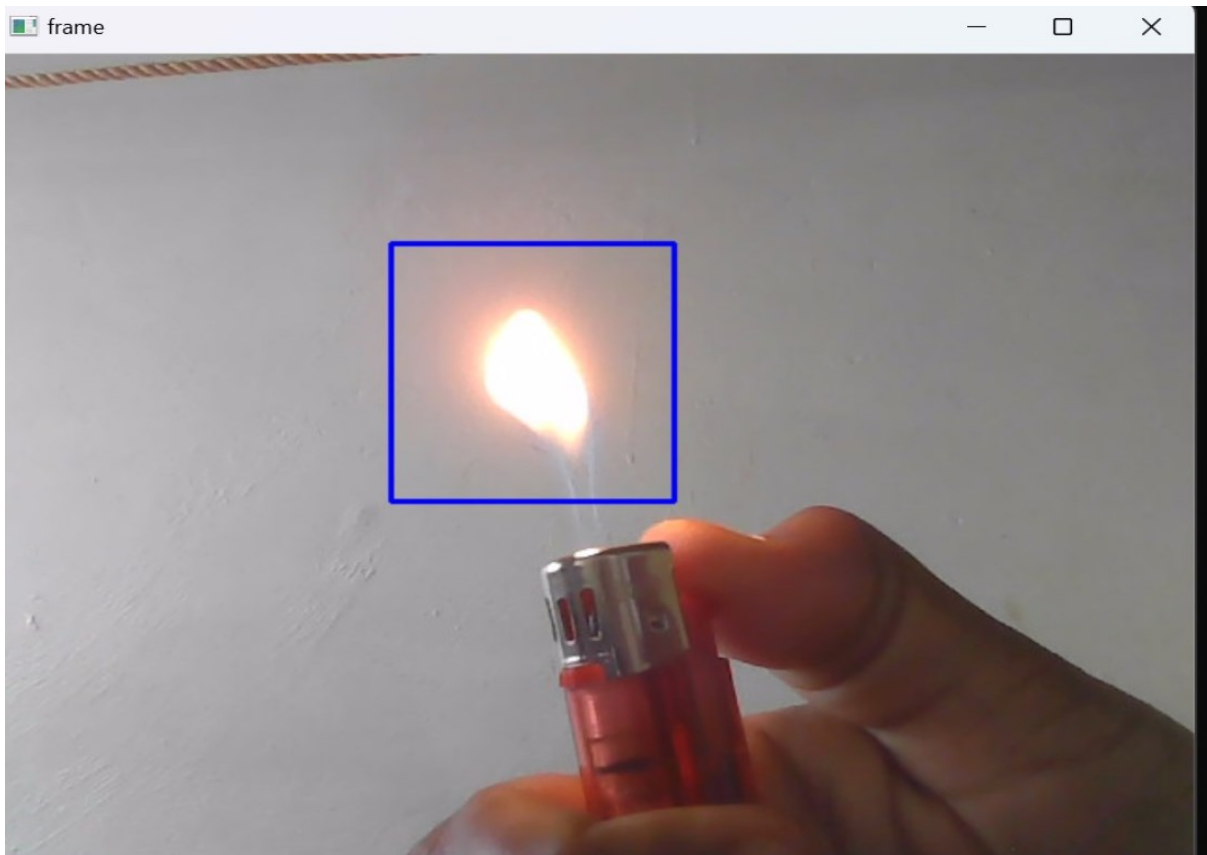


Fig 7:Fire Detected



Fig8:Fire Alert Initiated

5.CONCLUSION

An advancement in forest fire control, the forest fire response system based on deep

learning represents a significant step forward. The technology enhances detection accuracy, reaction speed, and

resource allocation with the use of deep learning algorithms and real-time monitoring. This study has the potential to enhance response operations by enhancing the safety of people, property, and the environment during forest fires.

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