

Skin Cancer Detection: A Survey¹

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ABSTRACT

Due to a lack of awareness of its warning signs and preventative measures, skin cancer—one of the deadliest types of cancer—has seen a significant increase in mortality rates. Therefore, early detection at an early stage is essential to halting the spread of cancer. Although there are other types of skin cancer, melanoma is the most dangerous. However, melanoma patients have a 96% survival rate when detected early with straightforward and cost-effective treatments. The project aims to classify various kinds of skin cancer using image processing and machine learning.

Melanoma is a type of skin cancer that can be fatal. If detected early, melanoma skin cancer can be completely treated. Because it directly correlates with death, early melanoma skin cancer detection is critical for patients. In this study, early melanoma skin cancer is detected and categorized using a variety of algorithms, including K-means clustering, neural networks, K-Nearest Neighbour, and Naive Bayes.

INTRODUCTION

Skin cancer is diagnosed in more people than any other type due to the depleted ozone layer and rapidly rising global air pollution. Compared to other types of skin cancer, melanoma has a much higher mortality rate. The cells that produce melanin, which is found in human skin, are known as melanocytes, according to research into the science of skin cancer. Melanoma skin cancer begins in melanocytes, which are the cells that produce melanin. Different melanocytes in the human body produce different amounts and types of melanin. It gives our skin color and shields it from the sun's ultraviolet rays. Having a family history of melanoma, having a lot or not enough moles, having certain skin types, and being in the sun for a long time are all risk factors for skin cancer. Although melanoma has a high mortality rate, there is a 99.99 percent chance of survival if detected early. Due to their striking similarities, dermatologists frequently have trouble distinguishing between benign and malignant lesions. There have been studies on how to diagnose melanoma skin cancer, but methods for more precise detection and classification are still needed. The K-means clustering method, neural networks, K-Nearest Neighbor, and Navie Bays are a few of the machine learning algorithms used in this paper to detect cancer. The

accuracy of the result is compared to these various classifiers.

A. Image Pre-Processing This identifies the image and removes all unwanted elements like noise, contrast, and hairs. They are removed, which may impact the method's accuracy, and a clear picture of the lesion is obtained so it can be easily identified. The following is a list of a few methods to eliminate this unwanted thing that could affect accuracy. The image below provides an overview of the various pre-processing tasks that can be carried out.

1) Enhancement of Images: Visual enhancement provides "better" contributes to other methods for automatic image processing and makes it easier for professionals to analyse image data. The fundamental objective of image enhancement is to modify an image's characteristics so that they are more suitable for a particular job or industry. There are numerous ways to enhance digital images without affecting them. These approaches are particularly problem-focused due to the close connection between the chosen parameters and the desired activity. The methods for improvement are best described by the following categories:

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a) methods in the frequency domain and b) techniques in the spatial domain

2) Grayscale to RGB conversion: Brightness is the only data in a grayscale image. Each pixel value represents a specific amount of light in a grayscale image. The brightness gradient can be distinguished in a grayscale image. In a grayscale image, only the intensity of the light is measured. Because grayscale images are quicker and easier to process than coloured images, the technology we propose converts color images into grayscale. After removing the hair and noise from the photos, we convert the noise-free ones to grayscale.

3) Noise-induced fluctuations in pixel value Our study employs the Non-local Mean Denoising method to remove undesirable elements from the skin image.) Eliminating Hair and Noise: The primary challenge in this study is distinguishing between essential features and noise-induced feat images.

4) Using a Gaussian Filter to Smooth: Gaussian smoothing distorts images. The Gaussian Standard Deviation measures the level of smoothing. The Gaussian filter neighbourhood average of each pixel, with the centre of the base pixels, weighted more heavily.

ALGORITHMS

KNN The k-nearest neighbours (KNN) method of supervised machine learning can deal with regression and classification issues. K-Nearest Neighbour (KNN) is the simplest, fastest, and most efficient classifier.

An image is categorized based on its neighbours' votes. The training and test samples are given to the KNN classifier, and the nearest distance is used to determine each class. 2) Nave Bayes: The prior probability conviction based on the serves as the primary foundation for a nave Bayes analysis.

The primary advantage is that only a little data is required. It is conditional, instantaneous independence in which no other qualities are required. The foundation of supervised machine learning is the naive Bayes method.[5] The Bayes theorem lays out the subjunctive possibilities that an occurrence x mentioning to a class k is feasible for identifying specific incidents in each kind of conditional probability as well as the unconditional possibility of the incident in each kind.

NEURAL NETWORKS

In artificial neural networks, a technique known as backpropagation is used to determine each neuron's contribution to the error after a batch of data—in image recognition, multiple images—has been processed. In such a circumstance, an enveloping optimization algorithm makes use of this to alter the weight of each neuron to complete the learning process. It is responsible for determining the loss function's gradient technically. The gradient descent optimization algorithm frequently makes use of it. 5]The fact that the error is calculated at the output and then distributed across the network layers is the source of the additional term "backward propagation of errors."

A. K-Means Clustering Algorithm

The K-means categorization algorithm is based on unsupervised learning. During k-means categorization, many clusters are created.

These clusters are only made up of a few data points. Each cluster divides the data points into many different categories. After the distinct clusters have been obtained, the accuracy is determined. The mean, median, standard deviation, minimum, variance, and maximum are used to derive the characteristics. These characteristics are provided as input to the k-means classification.

Here, two clusters are formed. The two clusters have the highest probabilities of skin pictures and the highest probabilities of cancer pictures, respectively. The K-means algorithm is utilized based on Euclidian distance measurements. There are two real data point clusters. At first, it is assumed that the cluster centres are selected randomly. The distance between the data points and the centroid is used in the calculation. Those clusters are made from data points whose distances to the centroid are the shortest. This procedure is repeated until the data points no longer move.

CONCLUSION

Melanoma is the most severe and aggressive type of skin cancer, so early detection is essential. You need an automated system to cut costs and make the process of finding melanoma more accurate. The advanced image processing technique described in this article uses a neural network to distinguish between nevus and melanoma. To segment images, the image segmentation technique is crucial. For image processing, the picture segmentation method is

essential. When melanoma skin cancer is discovered early, dermatologists can accurately and more quickly identify patients. Early location is fundamental since melanoma is the most dangerous and forceful skin malignant growth. To cut costs and improve detection accuracy, an automated melanoma detection system is required. We describe a method for finding skin lesions quickly. A cutting-edge image processing technique for separating melanoma from nevi. An artificial neural network (ANN) classifier can quickly determine the likelihood of a skin lesion developing cancer. As a result, we now know how limited feature extraction, segmentation, and pre-processing are when identifying skin lesions. To provide accurate results, the four stages of the melanoma diagnosis process use advance techniques. Melanoma in its early stages can be detected when these techniques are combined and applied to images of skin lesions.

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