



## Automated Skin Disease Prediction Using Deep Learning

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### ABSTRACT

Skin is an extraordinary human structure. He often suffered from many known and unknown ailments. Diagnosis of human skin diseases is therefore one of the most uncertain and complex areas of science. This article is dedicated to resolving this issue. Therefore, several experiments are performed on the dataset to evaluate the performance of the proposed system. This dataset consists of about 10,000 images collected from many sources such as Kaggle and various websites to make it more accurate and realistic. A comparative study was conducted on the application of different feature extraction algorithms using different classifiers. As a result, the name of the disease and the prescription associated with the treatment can be presented to the user with a high degree of accuracy, leading to the discovery of skin diseases.

**Keywords:** CNN; Skin disease; detection convolution Neural Networks.

### 1. INTRODUCTION

Skin diseases are often overlooked and receive little attention in their early stages. Awareness of individuals can contribute to the development of skin cancer. Currently, the skin irritation condition is only detected by biopsies at a future date. Examination is performed manually considering various histological features. Since this procedure is done manually, it can lead to human error; biopsy results are provided within 1-2 days.

In addition, the doctor has difficulty in determining the type of skin disease and its stage during the evaluation phase. As a result, prescribing medications is more complicated. This problem can be solved by evaluating the microscopic image using deep learning methods. This suggests that deep learning-based methods may be useful in quickly identifying clinical information and providing results. The multifaceted nature of skin disorders, the shortage and misallocation of competent physicians, and the critical requirement for rapid and accurate identification require informational treatment. Lasers and advances in photonics-based medical technology have enabled much faster and more accurate diagnosis of skin disorders. However, the costs of such diagnostics remain disproportionately high. Deep learning algorithms efficiently classify images and information.

### 2. LITERATURE

Several researchers have proposed photo processing-primarily based totally strategies to hit upon the kind of pores and skin sicknesses. Here we in brief assessment a number of the strategies as mentioned within side the literature.

In [1], In the approach of [2], extraction of photo functions is step one in detection of pores and skin sicknesses. In this approach, the extra range of functions extracted from the photo, higher the accuracy of device.

The creator of [2] implemented the approach to 9 kinds of pores and skin sicknesses with accuracy as much as 90%.

Melanoma is kind of pores and skin most cancers that may purpose death, if now no longer diagnose and deal with within side the early stages. The creator of [3], targeted at the take a look at of diverse segmentation strategies that would be implemented to hit upon cancer the use of photo processing. Segmentation system is defined that falls at the inflamed spot barriers to extract greater functions.

The paintings of [4] proposed the improvement of a Melanoma analysis device for darkish pores and skin the use of specialized set of rules databases consisting of pictures from a whole lot of Melanoma resources. Similarly, [5] mentioned category of pores and skin sicknesses which includes Melanoma, Basal mobileular carcinoma (BCC), Nevus and Seborrheic keratosis (SK) with the aid of using the use of the method aid vector system (SVM). It yields the high-quality accuracy from quite a number of different strategies.

On the opposite hand, the unfold of continual pores and skin sicknesses in special areas might also additionally result in excessive consequences. Therefore, [6] proposed a laptop device that routinely detects eczema and determines its severity. The device includes 3 stages, the primary powerful

segmentation with the aid of using detecting the pores and skin, the second one extract a fixed of functions, specifically shadeation, texture, borders and 0.33 decide the severity of eczema the use of Support Vector Machine (SVM).

In [7], a brand-new method is proposed to hit upon pores and skin sicknesses, which mixes laptop imaginative and prescient with system mastering. The position of laptop imaginative and prescient is to extract the functions from the photo whilst the system mastering is used to hit upon pores and skin sicknesses. The device became examined on six kinds of pores and skin sicknesses with correctly 92%.

### 3. Description

Skin disorders include those that originate in an organ and originate in the skin, or those that originate and originate. This section analyses these diseases. We also describe the topology of an artificial neural network used for diagnosing skin diseases. The patient's information and symptoms are collected, and a specialist diagnoses the skin condition. This collection of documented skin diseases and conditions has since been expanded to include a number of logical indicators. If a result is obtained, the doctor will suggest that condition as the cause of the skin condition. In some cases, the specialist may refer the patient for further laboratory tests to determine the etiology of the skin condition. This test can be performed to confirm that the disease being evaluated is caused by organisms such as bacteria, parasites, viruses, or fungi [26]. A specialist who is unskilled or who has never seen such skin conditions diagnoses by trial and error. This is achieved by combining all possible situations and comparing them to known situations to narrow down the decision. Learning is believed to have taken place at this stage when skin problems are properly diagnosed and treated. Experts therefore rely on their own expertise and assessment of customer complaints.

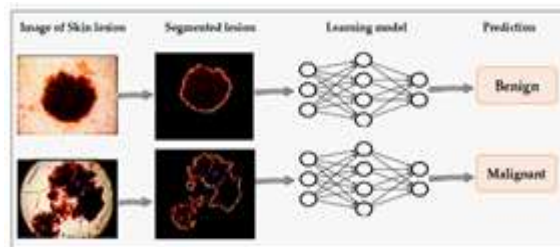


Fig3.1. skin cancer detection

### 4. Methodology

All n this section, the method of the proposed machine for detection, extraction and class of pores and skin sicknesses pics is described. The machine will assist notably within side the detection of melanoma, Eczema and Psoriasis. The complete structure may be divided into numerous modules comprising of preprocessing, function extraction, and class. The block diagram of the machine is proven in Fig 2

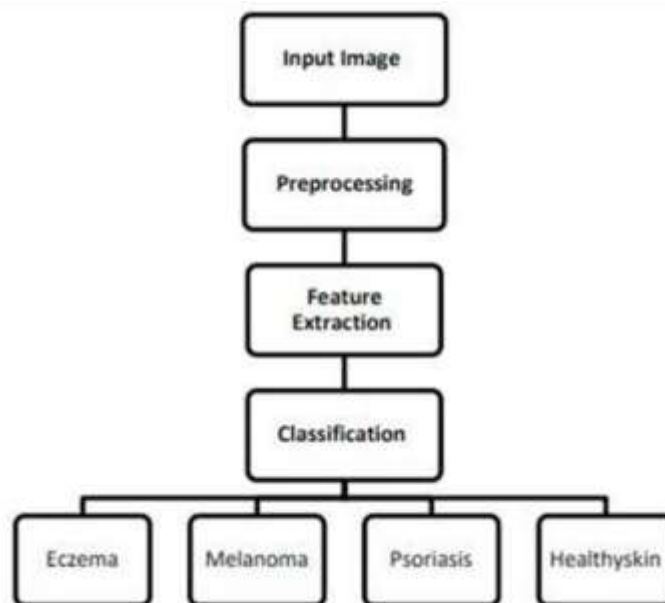


Fig. 4.1. The proposed system block diagram

#### 4.1 Preprocessing

Achieving excessive overall performance of pores and skin sickness detection gadget calls for overcoming a few important difficulties. Such as growing a database and unifying photo dimensions. In the subsequent section, the approach utilized in photo resizing is explained.

#### 4.2 Image Resizing

To clear up the trouble of various photo sizes with inside the database an enter photo is both boom or lower in length. Unifying the photo length gets the equal variety of capabilities from all images. Moreover, resizing the photo reduces processing time and as a consequence will increase device performance. Fig three suggests the authentic photo of length is 260×325 pixels. Fig four suggests the resized photo with the brand-new length of 227×227 pixels.



Fig. 4.2. Example of Original image of Eczema database.



Fig. 4.3. Example of resizing image of Eczema database.

#### 4.3 Feature Extraction

At the beginning, Convolutional Neural Network (CNN) is a fixed of stacked layers regarding each nonlinear and linear processes. These layers are discovered in a joint manner. The primary constructing blocks of any CNN version are: convolutional layer, pooling layer, nonlinear Rectified Linear Units (ReLU) layer linked to a normal multilayer neural community known as absolutely linked layer, and a loss layer on the backend. CNN has acknowledged for its massive overall performance in packages because the visible responsibilities and herbal language processing.

#### 4.4 Classification

Classification is a laptop imaginative and prescient method. After extracting capabilities, the position of class is to stylish the photograph thru Support Vector Machine (SVM). A SVM can educate classifier the use of extracted capabilities from the schooling set .

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## 5. Requirements

### 5.1 Software requirements

- Jupyter Notebook
- Python version
- SQLite database

- MATLAB 2018b

### 5.2 Hardware requirements

- Intel Core i3
- processor 2.10GHz
- RAM 4GB

## 6. Proposed Solution

### WHAT IS TO BE DEVELOPED

For detecting the Skin disease prediction using image processing and CNN algorithm.

### PROPOSED SYSTEM

In proposed system we are developing a Web application. A backpropagation neural network (Neural Works Professional II/Plus) was used to analyze images of patients with skin disease and images of healthy subjects. Using the Kaggle database, the feature extraction method and CNN were efficiently processed to determine the presence or absence of skin diseases.

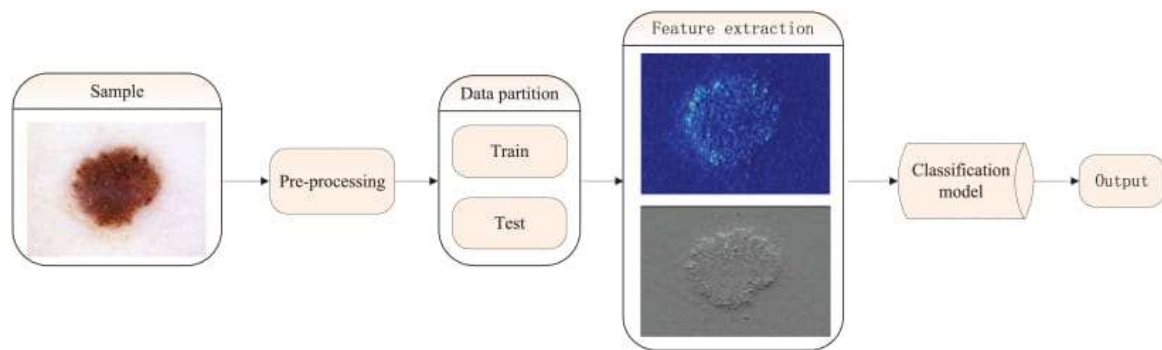


Figure 6.1.-: Flowchart of proposed system

### CONVOLUTIONAL NEURAL NETWORKS

#### Why Convolutional Neural Networks Are So Important?

Convolutional networks are important to learn for image and computer vision tasks. Convolutional networks have the property of "spatial invariance". That is, it learns to recognize image features everywhere in the image.

So, how do convolutional neural networks actually operate?

The first thing we need to know here is the elements involved in the operation:

- Input image
- Convolutional Neural Network
- Output label (image class)

This element interacts in the following manner:

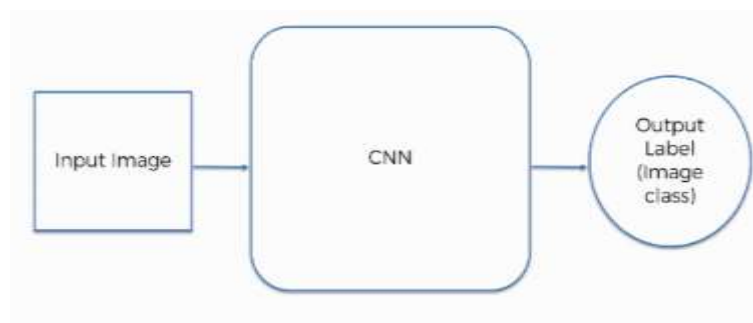


Figure 6.2.-: CNN Elements

For example, convolutional neural networks can be used when human emotion is detected in an image. Give someone a picture and we'll categorize how that person is feeling. Of course, this requires a bit of advanced training, as it is often very difficult to infer people's emotions from facial expressions. Considerable success rate.

How Convolutional Neural Networks Scan Images

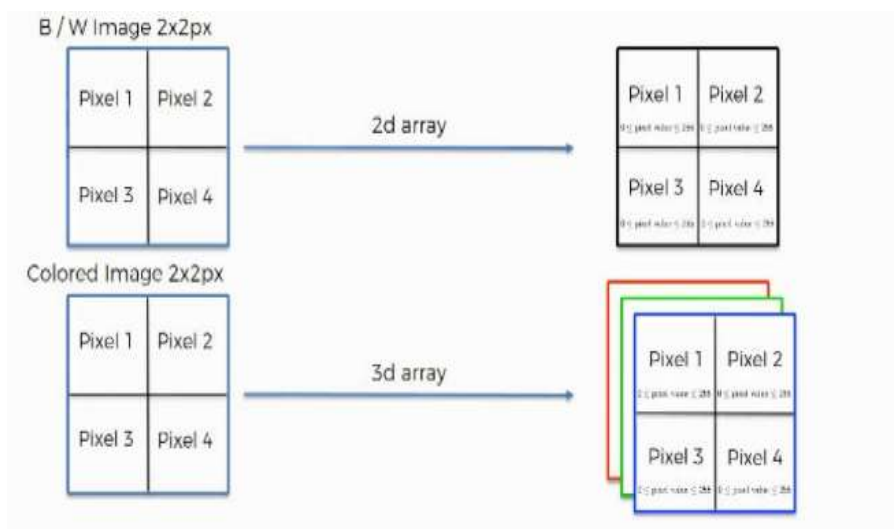


Figure 6.3.- CNN Image Scan Process

Both image types are similar in that:

- Each pixel contains 8 bits of information.
- Colours are represented on a scale of 0-255. The reason for this is that bits are binary units, there are 8 per byte, so a byte can have one of 256 ( $2^8$ ) possible values. Count 0 as the first possible value, so you can go all the way to 255.
- In this model, 0 is pure black and 255 is pure white, with various (well over 50!) shades of gray in between.
- The network doesn't actually learn colours. Computers only understand 1's and 0's, so colour numbers are represented in the network in binary form.

Let's discuss the main differences mentioned above. The black-and-white images are two-dimensional, while the colour images are three-dimensional. The difference is in the value assigned to each pixel when presented to the neural network. In a black and white two-dimensional image, each pixel is assigned a number between 0 and 255 to represent its hue.

On the other hand, each pixel in a colour image is represented in three levels. Each colour is a combination of red, green, and blue with different densities, so a single pixel in his colour image is assigned a separate value for each of these layers.

This means that the red layer is represented by a number between 0 and 255, just like the blue and green layers. It is then displayed in RGB format.

## 7. Application

- To extend the role of the technology in the medical field.
- Any medical related survey sections.
- The complex medical system, which will facilitate the development of medical
- Diagnosis and clinical application as well as promote the development of the medical field.

## 8. Conclusion

In this work, a model is built for predicting skin diseases using deep learning algorithms. It is found that by using ensemble features and deep learning, we can aim to achieve higher accuracy and predict more diseases than any other model done before. Similar to previous models in this application area, we were able to report up to nine skin disorders with maximum accuracy. By implementing deep learning algorithms, we are now able to predict a large number of diseases more accurately. This proves that deep learning algorithms have great potential for diagnosing skin diseases in the real world. Using a better system in high-end system hardware and software with very large data sets would greatly improve accuracy and the lack of invasive means would allow the model to be used in clinical experiments. This model can be extended in future work to make it a standard procedure for pre-diagnostic methods of skin diseases, as it shortens the time of treatment and diagnosis.

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