



Skin Cancer Disease Detection Using Deep Learning

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Abstract— Skin cancer is among the world's most hazardous illnesses. Skin cancer is challenging to accurately diagnose, though. Recently, deep learning systems performed a variety of tasks exceptionally well. Additionally, they have been employed expressly for tasks including the detection of skin diseases. The use of deep learning algorithms to identify skin disorders is examined in this article. We start by introducing several publicly available datasets for developing and evaluating algorithms for the detection of skin disorders, image collection methods, and dermatology. Then, we introduce the idea of deep learning and examine popular deep learning architectures. In the future, we can look into it. (Abstract)

Keywords- skin conditions, Model selection, Model training, Model evaluation, Model tuning, Deployment, Continuous improvement, medical diagnosis, Dermatology, Convolutional neural network, Artificial intelligence, Deep learning, Computer vision

1. Introduction

A common and challenging condition to diagnose due to ignorance and lack of understanding is skin disease. People consult dermatologists for skin diseases and preventative

measures in many developing nations. The present system is without reason, and the public is unsure about the dermatologists' medication recommendations. The importance of treating skin conditions early on cannot be overstated, since the skin performs a critical defense function for the body against dangerous bacterial and fungal diseases. Due to their genes, jobs, poor diets, typical settings, exposure to chemicals, etc., many people develop skin issues. Skin disease is influenced by environmental variables such as climate, summer season, and winter season less time. the use of photography techniques, the suitable skin condition is detected, with a proposed diagnosis. The input analysis is done in two steps to address this problem. To train the model, there are two methods: one uses image processing, the other uses machine learning. This model is always being enhanced to predict different skin disorders. It is necessary to teach the machine learning algorithm for accurate prediction thus different skin disorders have different characteristics. Skin problems are routinely ignored and given little thought in the early stages. Skin cancer could result from the ignorance of humans. The expanded skin illnesses are just distinguished through biopsy in the current procedure at a later stage. In the existing method, the increased skin illnesses are only discovered later by biopsy.

II. Related Work

To identify the many types of skin diseases, several researchers have suggested image processing-based approaches. In this section, we discuss a few of the techniques in the literature that have been characterized. In, a technique is proposed for analyzing skin disorders using colour photos without the aid of a doctor. The technique consists of two steps: the first includes recognizing skin disorders using approaches for colour gradients, k-means clustering, and colour image processing, and the second involves determining the kind of illness using artificial neural networks. The accuracy averages for the first stage and second stage of the approach were 95.99% and 94.016%, respectively, when it was examined on six different kinds of skin conditions. The first stage of the feature extraction procedure for images. With up to 90% accuracy, The technique was used on nine distinct skin disorders by the author of. the method was developed by and was applied by nine distinct skin illnesses with up to 90% accuracy.. Skin cancer known as melanoma can prove fatal if When it first appears, it is not identified and treated. The author's primary area of interest was the analysis of various segmentation methods for identifying melanoma that make use of image processing. It is described how to segment data using a sick spot's borders to get more details. The study suggested creating a tool for diagnosing melanoma in people with dark skin utilizing specialized databases of algorithms that also contained images from other melanoma websites. Similar to this, support vector machine (SVM) was used to categories skin conditions such seborrheic keratosis (SK), basal cell carcinoma (BCC), nevus, and melanoma. From a variety of different procedures, it produces the most accurate results. On the other side, the spread of chronic skin conditions beyond geographical boundaries might have negative effects that are severe. Consequently, a computer system that automatically recognizes eczema and assesses its severity was proposed. Using Support Vector Machines (SVM), determine the severity of eczema. In, a novel method for identifying skin illnesses is put forth that combines computer vision and machine learning. While computer vision is used to extract features from photos, machine learning is used to identify skin issues. Six different skin conditions were evaluated, and the system performed 95% correctly.

III. Literature Survey

[3] In this paper, The authors suggested a model to analyse input skin photos for skin lesions and identify them using medical imaging. They developed a prototype system to find skin diseases using approach. The goal of this research is to identify skin lesions using thresholding and neural network-based texture analysis of the input skin photos to identify and diagnose skin illness.

[4] In this research paper, describes a system for forecasting skin disease based on user-provided input photographs and

responses to a set of questions. The system uses a trained model to identify the kind of skin condition and provides recommendations for medical treatment. However, the paper notes that the system may not always yield positive outcomes.

It is significant to emphasise that the system's accuracy predictions may depend on the quality of the user-provided input photographs and the user's responses to the questions. It is also possible that the trained model may not be able to accurately identify certain types of skin diseases. This study examines skin disorders include urticaria, eczema, and fungal infections. This application's reliance on questions and responses doesn't always result in fruitful results.

IV. SYSTEM ANALYSIS

A. EXISTING SYSTEM:

Now, people live longer. In actuality, both the greatest observed age of death and life expectancy have greatly increased in recent decades. It is currently not feasible, beneficial, or moral to merely alter human DNA or restrict food intake in order to extend a healthy life. The extension of life is one of the biggest concerns facing society in the twenty-first century. The greatest possible life duration for humans is still a hotly contested topic. Many scientists agree that there is an innate upper limit to human life, although they disagree as to whether it is 85, 100, or 150. The average human life expectancy is believed to be roughly 125 years, with death rates increasing in elder age nowadays.

B. PROPOSED SYSTEM:

The dataset is first gathered when the procedure begins. Data preprocessing, which comes after data filtration, data reduction, and data transformation, is the next stage. Following that, the skin condition will be forecasted Convolutional Neural Network (CNN) technology is being used. the convolutional neural network algorithm. Based on input from users, the model will forecast the skin illness after utilising the algorithm. Through the website, users may contribute to the machine learning prediction of skin diseases.

C. PROBLEM STATEMENT:

The prototype receives as input from the patient a picture of the skin region that is diseased. This photograph is subjected to image processing tools, and the result shows the disease that was discovered.

D. STATEMENT OF SCOPE

The biggest organ in the human body, skin covers the bones, protects against injury, wards off disease-causing microorganisms, and has a wide range of possible disorders. Numerous variables can directly or indirectly impact the skin and result in disorders that can be treated with particular medications while others call for a doctor's visit.

By processing the image, extracting information that reveals the area of the skin that is afflicted, and categorizing the image according to the kind of skin sickness, this article will assist people in understanding the methods necessary for treating skin diseases.

E. GOAL AND OBJECTIVES:

The project's goal is to quickly and accurately identify the kind of skin illness and provide the appropriate treatment. Before feature extraction, several pre-processing procedures are used to the skin disease in the picture. The second stage includes analysing and observing the skin, followed by using machine learning algorithms to diagnose illnesses. The proposed system is really helpful in rural locations with limited dermatologist access. We utilise a Python script built on Pycharm to gather data for the suggested system.

SYSTEM ARCHITECTURE:

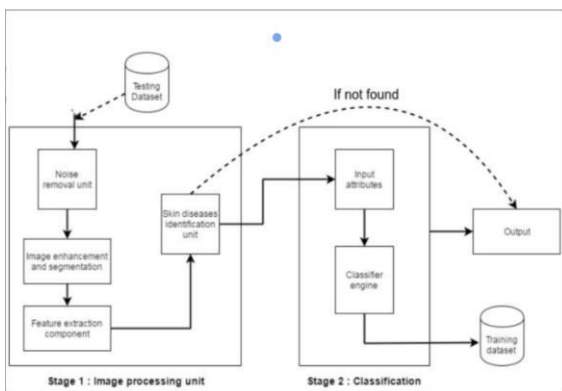


Fig 1. System Architecture

V. MOTIVATION OF THE PROJECT

The most prevalent disease in the world is skin disease. One primary motivation for developing a computer-aided skin disease diagnostic model is to improve the accuracy and objectivity of skin disease diagnosis. Traditional methods of skin disease diagnosis often rely on visual inspection and interpretation by a dermatologist, which can be subjective and prone to errors. By using a computer-aided diagnostic model, the hope is to reduce the likelihood of misdiagnosis and provide more consistent and reliable diagnoses. By leveraging advancements in computer vision and machine learning, researchers and developers are able to create innovative solutions that have the potential to revolutionize medical diagnosis and treatment.

VI. PROPOSED METHODOLOGY

Detecting skin diseases using deep learning is a promising

area of research that has the potential to improve diagnostic accuracy and speed up the process of diagnosis. Here is a proposed methodology for skin disease detection using deep learning:

Data collection: Collect a large dataset of images of skin diseases. This dataset should include a wide range of skin conditions, including common diseases like acne, eczema, and psoriasis, as well as rare diseases.

Data preprocessing: Before feeding the data into the deep learning model, it needs to be preprocessed. This can include image resizing, normalization, and augmentation to increase the variety of data.

Model selection: Select an appropriate deep learning model for skin disease detection. The model should be able to handle large datasets and be capable of detecting subtle differences in skin textures and colors.

Model training: Train the deep learning model using the preprocessed data. This involves feeding the model with the dataset and adjusting the model's parameters to optimize performance.

Model evaluation: Evaluate the trained model's performance on a test dataset. This will help to determine how well the model performs on unseen data.

Model tuning: Based on the evaluation results, adjust the model's parameters to improve performance.

Deployment: Deploy the trained model in a real-world application. This can include developing a mobile application or web-based tool for dermatologists and healthcare professionals to use for skin disease diagnosis.

Continuous improvement: Monitor the performance of the model in real-world settings and continue to make improvements to the model over time. This can include adding new data to the dataset and refining the model's parameters to improve accuracy and speed.

Overall, Deep learning has the ability to greatly increase diagnostic precision and speed up the diagnosis process for skin diseases.. However, it is important to carefully select and train the deep learning model and continually improve its performance over time.

VII. RESULTS

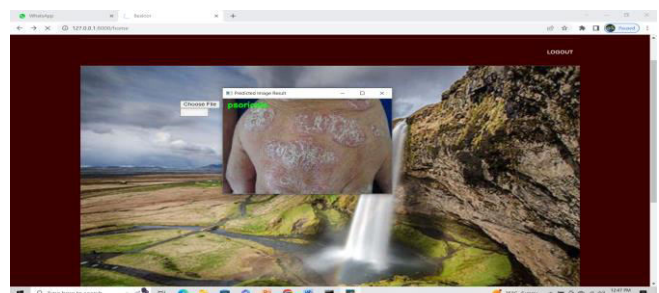


Fig 2. Classification Results

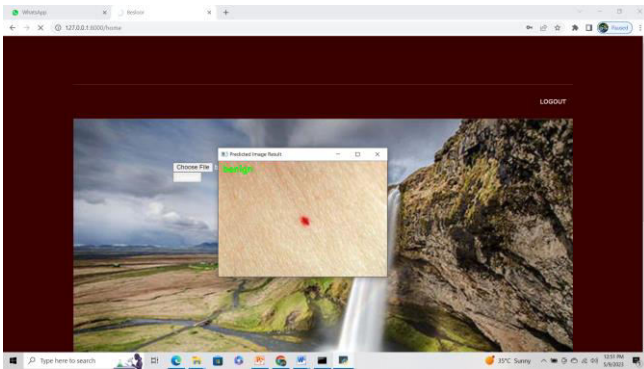


Fig 3. Classification Results

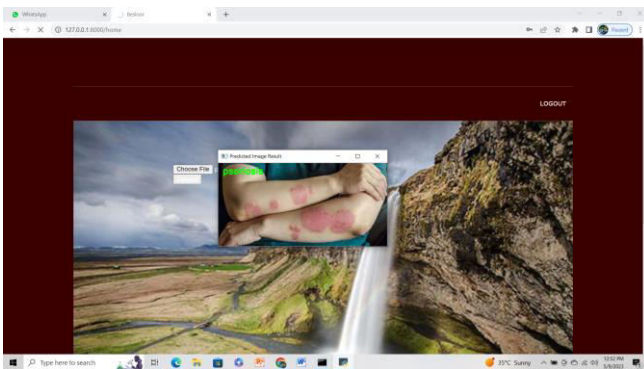


Fig 4. Classification Results

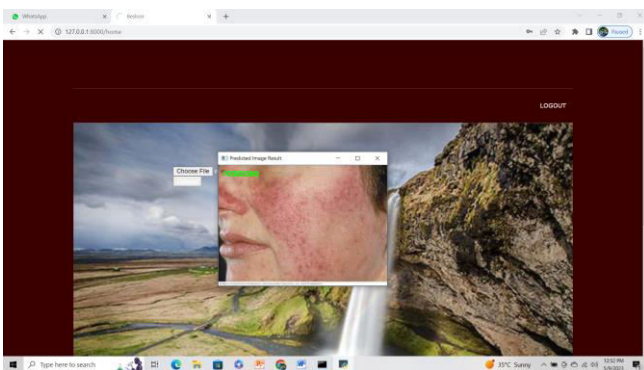


Fig 5. Classification Results

VIII. CONCLUSION

In this study, A skin disease prediction model is developed using deep learning methods. Researchers have shown that by using deep learning and ensembling traits, they can predict more illnesses with more accuracy than they could with any other model up to that point. as the previous models created for this application could only report a maximum of two skin problems with a 75% accuracy level. By using a deep learning system, we can predict up to 4 illnesses with an accuracy rate of 88%. This demonstrates that deep learning algorithms have a great deal of potential for real-world skin condition diagnosis. This model may be used for clinical research given that it doesn't require any

invasive procedures, and Additionally, the accuracy enhanced by employing still more sophisticated tools a very huge dataset, and software. As it will speed up therapy and diagnosis, further study may be carried out to standardise this concept. as a strategy for identifying skin diseases early on.

IX. FUTURE SCOPE

Future connections between our machine learning model and numerous It is feasible to find websites that provide real-time data for predicting skin diseases. We could also include a lot of historical data on skin diseases, In order to increase the accuracy of the machine learning model. As a user interface for engaging with the user, we can create an Android app. To optimise performance, we wish to carefully design the topologies of deep learning networks, Instead of training the networks on the entire dataset, utilise adaptive learning rates to train the networks on data clusters.

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