



# Deep Learning Assisted Skin Cancer Detection using Smart Healthcare Web Application

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## ABSTRACT

*Skin cancer is a major worldwide health issue that requires timely and precise diagnosis to enhance patient outcomes. Although the clinical evaluation of skin lesions is essential, it frequently faces obstacles such as extended waiting periods and subjective judgments. Deep learning algorithms provide effective answers to these difficulties by aiding dermatologists in making more accurate diagnoses. Deep learning algorithms can improve the speed and accuracy of diagnosis, resulting in earlier discovery and beginning of therapy. Furthermore, these algorithms have the potential to reduce the workload of healthcare practitioners, enabling them to concentrate on more intricate situations. The objective of this project is to create strong and reliable deep learning models that can accurately classify skin cancer. This study will specifically address challenges such as imbalanced classes and will also provide valuable insights into the decision-making process. In addition, the study suggests the implementation of a comprehensive smart healthcare system using an android application. By integrating attention mechanisms and visualization approaches into deep learning models, interpretability issues are resolved, allowing doctors to acquire understanding of the fundamental features of skin lesions and assisting in decision-making procedures. This application highlights the potential of deep learning to enhance the accuracy and accessibility of diagnostic procedures in dermatological healthcare.*

**KEYWORDS:** Image Processing, skin cancer classification, deep learning, diagnostic accuracy, dermatological healthcare

## 1. INTRODUCTION

A problem that is quite common and has important repercussions for the health of the general population is skin cancer, which is defined by aberrant cell development in the epidermis as a result of DNA damage. A significant contribution to the success of

therapy and the enhancement of patient outcomes is made by early identification. The burden of skin cancer continues to be enormous, despite the gains that have been made in medical research. As a result, novel techniques are required to address the obstacles that are associated with diagnosis. The purpose of this project is

to improve the diagnosis of skin cancer by utilizing the power of deep learning algorithms. The goal is to provide a solution that is both scalable and effective for both medical professionals and patients. Through the utilization of the capabilities of artificial intelligence, the system is able to perform skin lesions analysis with a high degree of accuracy and speed. This allows for the early identification of suspicious growths and affords the opportunity for prompt action.

The construction of a robust deep learning model that has been trained on huge datasets of skin photos that have been tagged with diagnostic labels is one of the most important goals of the research. It is possible for the model to acquire the ability to differentiate between benign and malignant lesions through iterative refinement and validation, which will assist dermatologists in exercising informed clinical decision-making. To add insult to injury, the incorporation of cutting-edge technology into an intuitive user interface guarantees accessibility and usefulness for both patients and healthcare practitioners. Individuals are given the ability to proactively monitor their skin health and seek medical assistance when it is required by the system, which does this by providing the data in a manner that is both plain and understandable.

Furthermore, the suggested system offers a comprehensive approach to the treatment of skin health, which is a departure from the conventional diagnostic tools that are now available. Users are able to obtain individualized information on preventative actions and lifestyle adjustments to minimize their chance of acquiring skin cancer through the use of features such as risk assessment and personalized suggestions. In addition, the scalability of the system makes it possible to integrate it without any disruption into the current healthcare infrastructure, which makes it easier for the system to be adopted and implemented across a wide range of patient groups.

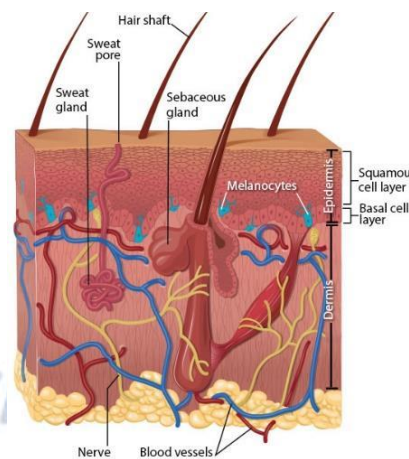


Figure 1. Skin layers

As a conclusion, this study marks a significant step forward in the utilization of technology to meet the issues connected with the identification and management of skin cancer. The suggested system provides a solution that is both scalable and efficient for early detection and intervention. This is accomplished by leveraging the power of deep learning algorithms and integrating them into an interface that is user-friendly. This project has the ability to enhance patient outcomes and minimize the strain on healthcare systems, which shows that it has the potential to advance public health efforts in the battle against skin cancer.

## 2. LITERATURE SURVEY

Discovering the promising potential of artificial intelligence (AI) in transforming skin cancer diagnostics has been made possible by research conducted in the fields of dermatology and deep learning. Numerous research have emphasized the efficiency of Convolutional Neural Networks (CNNs) and other deep learning models in correctly detecting skin lesions and discriminating between benign and malignant cancers. As an illustration, Esteva et al. (2017) proved that a deep learning system is capable of accurately diagnosing skin lesions, with a level of performance that is comparable to that of highly trained dermatologists. In a similar vein, Tschandl et al. (2019) shown that CNNs are capable of effectively detecting melanoma, reaching high sensitivity and specificity rates in the process. In addition, Haenssle et al. (2018) carried out a groundbreaking study in which they compared the performance of a CNN-based algorithm with that of dermatologists, and they found that both methods had equivalent diagnostic accuracy respectively.



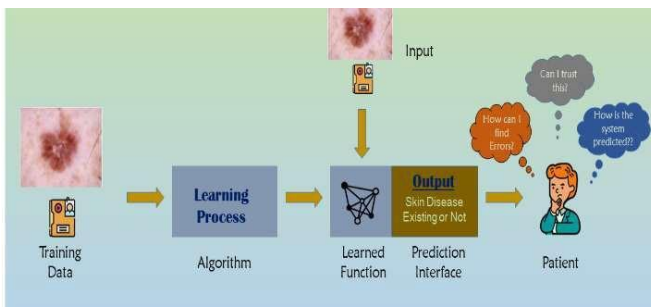


Figure 2. Skin cancer detection

Researchers have made tremendous progress in the field of automated skin cancer diagnosis by utilizing powerful algorithms and utilizing enormous datasets that contain annotated photos of the skin. Through the application of deep learning strategies, the creation of diagnostic tools that are both more effective and more trustworthy in the medical field has been enabled. For example, Brinker et al. (2019) introduced a deep learning model that is capable of identifying skin lesions based on dermoscopic pictures. This demonstrates the potential of the model to improve early diagnosis and reduce the number of needless biopsies. In addition, Menegola et al. (2020) investigated the possibility of incorporating machine learning algorithms into dermoscopic imaging systems in order to enhance the precision of melanoma detection.

Collectively, these works provide light on the revolutionary influence that artificial intelligence has had on the field of dermatology, notably in the field of skin cancer diagnosis. Researchers continue to make progress in the creation of AI-driven diagnostic systems by utilizing cutting-edge technology and extensive datasets. This helps to provide better outcomes for patients and improves the efficiency with which healthcare is delivered.

### 3. EXISTING SYSTEM

The dependence on conventional diagnostic procedures for the identification of skin cancer presents a number of important issues within the context of the healthcare system. Despite the fact that manual inspection of skin lesions is regarded to be the gold standard, it is fundamentally subjective and vulnerable to variations among medical practitioners. It is possible that this subjectivity will result in contradictions in diagnosis, which might potentially postpone the beginning of therapeutic treatments at the appropriate moment. Furthermore, the procedure of manual

inspection is frequently time-consuming, which places a burden on the resources available in the healthcare industry and contributes to lengthier wait times for patients who are seeking diagnosis and treatment.

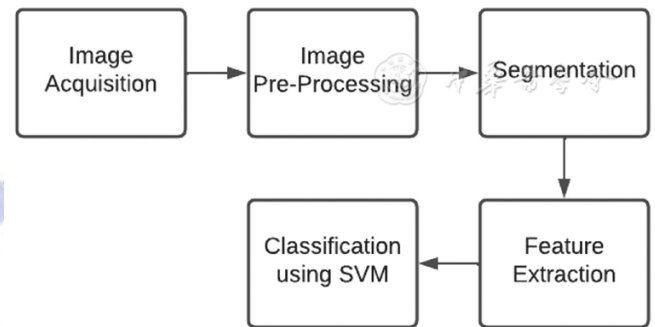


Figure 3. Existing model

In addition, the availability of professional dermatologists and resources for the detection of skin cancer might be restricted, particularly in locations that are underserved or rural. It is possible that chances for early identification and intervention may be lost as a result of this lack of access, which compounds existing discrepancies in the delivery of healthcare. It is possible that patients in these areas may have a difficult time gaining access to diagnostic evaluations that are both quick and accurate, which will in turn reduce the likelihood that they will experience positive outcomes from their therapy.

As a result of these issues, there is an urgent requirement for the development of novel technologies that are capable of enhancing the diagnostic procedures that are already in place and overcoming the constraints of the conventional approaches. When it comes to this matter, artificial intelligence (AI) and machine learning (ML) algorithms provide some possibilities for potential answers. Systems that are driven by artificial intelligence are able to evaluate and understand complex patterns and characteristics that are symptomatic of skin cancer with a high level of accuracy and efficiency. This is accomplished by using vast databases of skin scans and related clinical data.

Because they are capable of providing automated, objective, and speedy diagnostic capabilities, these cutting-edge technologies have the potential to completely transform the process of detecting skin cancer. Systems that are driven by artificial intelligence can give assistance to healthcare practitioners by identifying worrisome lesions that require additional assessment. This helps to streamline the diagnostic

workflow and reduce the amount of work that physicians have to do. Furthermore, the incorporation of diagnostic tools driven by artificial intelligence into clinical practice offers the potential to enhance diagnostic accuracy and consistency across a wide range of healthcare settings.

Artificial intelligence-based systems have the potential to democratize access to skin cancer detection by eliminating geographical obstacles and expanding the availability of knowledge. This is in addition to the fact that these systems have the capacity to improve diagnostic skills. Telemedicine systems that are integrated with diagnostic tools driven by artificial intelligence can offer remote consultations and triage, which enables patients in underserved regions to obtain fast evaluations and recommendations for additional examination as required.

In general, the creation and use of cutting-edge technologies for the identification of skin cancer constitute a big step forward in terms of enhancing the results for patients and resolving inequities in healthcare. By leveraging the power of artificial intelligence and machine learning, medical professionals may improve the speed, accuracy, and accessibility of skin cancer detection, which will eventually result in the saving of lives and an improvement in the quality of treatment for patients all over the world.

#### 4. PROPOSED SYSTEM

By using deep learning models trained on a large and diverse dataset of skin lesion photos, the suggested method advances skin cancer diagnosis. This revolutionary approach uses artificial intelligence to improve diagnosis accuracy and deliver transparent forecasts, unlike manual inspection and subjective interpretation. The system relies on cutting-edge deep learning architectures including CNN, Efficient Net, and InceptionV3. These advanced models have been rigorously trained on a broad array of skin lesion photos to learn intricate skin cancer patterns and traits. Using ensemble learning to integrate the benefits of many models improves sensitivity, specificity, and diagnostic accuracy.

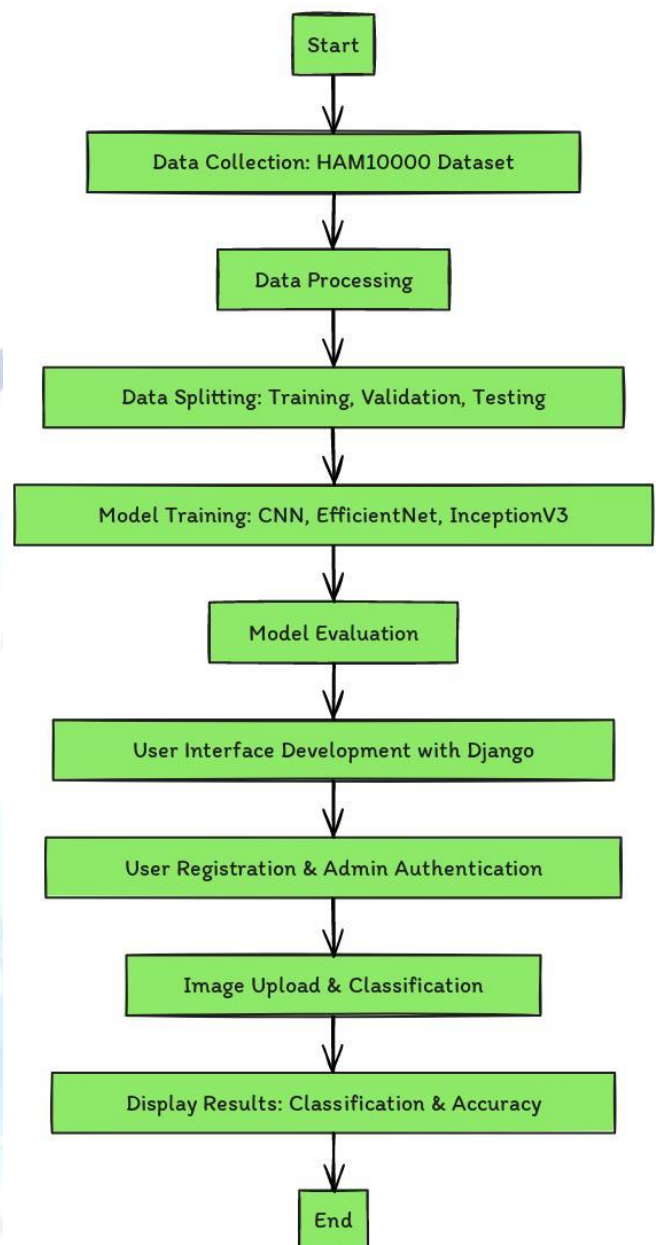


Figure 4. Proposed model

The Django-based system's user-friendly web interface is a crucial feature. This interface makes system interaction secure and easy for healthcare professionals and patients. Users may input skin lesion photographs, obtain real-time diagnostic forecasts, and view AI-driven decision explanations via the online interface. The technology builds user trust and empowers them to make educated patient care decisions by making the diagnostic process visible and interpretable. The suggested approach also stresses data privacy and ethics, protecting sensitive patient data and complying with regulations. The system protects patient confidentiality and medical ethics by following strict data protection and ethical criteria. The suggested



technology smoothly fits into healthcare operations, complementing standard diagnostic methods and improving access to sophisticated dermatological diagnostics. Physicians may use the system as a decision support tool to improve diagnosis and patient outcomes. The technology might also provide sophisticated dermatological diagnostics to underprivileged and distant populations, eliminating healthcare inequities and boosting health equity. A powerful and scalable solution that blends cutting-edge technology with user-centric design, the suggested system advances skin cancer diagnosis. The technology might transform skin cancer diagnosis and management by using deep learning and offering transparent, interpretable diagnostic predictions, saving lives and improving treatment globally.

## 5. IMPLEMENTATION

Creating and curating a large dataset of skin lesion photos from various skin cancers is the first step in implementing the suggested method. Deep learning algorithms learn skin condition-specific features and patterns from this dataset. The system uses state-of-the-art architectures like Convolutional Neural Networks (CNNs) to extract and abstract key information from input photos to build a robust classifier that can properly classify skin lesions. Scalability, usability, and diagnostic prediction transparency are key to the system's architecture. Optimization techniques iteratively alter model parameters to decrease prediction errors and increase accuracy and sensitivity in deep learning models. The system optimises its algorithms across varied datasets and clinical circumstances through repeated testing and validation. The solution includes model training and interpretability tools to help users understand the AI-driven diagnostic system's decision-making process. Healthcare practitioners may better comprehend and interpret diagnostic predictions by displaying the deep learning models' prominent characteristics and regions of interest, boosting their trust in the system's suggestions.

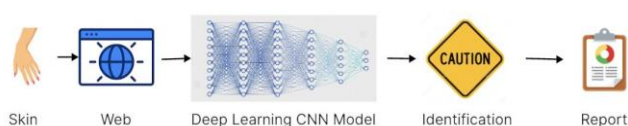


Figure 5. Deploying model of the project

The technology also protects patient data from illegal access and exploitation by following strict data privacy and encryption standards. The proposed skin cancer detection system addresses current skin cancer detection system constraints including diagnosis variability and restricted access to specialist knowledge to improve dermatological diagnostics reliability and accessibility. The system's user-friendly interface and transparent diagnostic forecasts allow healthcare practitioners to make educated patient care decisions, increasing outcomes and lowering resource strain. The suggested technology might transform dermatology by enabling early skin cancer diagnosis and intervention, saving lives and enhancing patient care globally.

## 6. RESULTS

The provided text outlines the functionality and security measures implemented in a skin cancer detection web application. The application employs end-to-end encryption to ensure the confidentiality and integrity of user data throughout the entire process, from image upload to analysis results. This encryption protocol prevents unauthorized access or interception of sensitive health information, providing users with peace of mind regarding their privacy and data security. Additionally, the application features a user-friendly dashboard designed to streamline the user experience and provide easy access to essential functionalities. Users can manage their accounts, track their skin health journey, and access valuable insights from the dashboard. Moreover, the text emphasizes the importance of providing clear and detailed descriptions when uploading images for skin cancer detection. Relevant patient information such as age, gender, medical history, and known risk factors for skin cancer should be included to facilitate accurate analysis by the detection system. The description of a skin lesion is also provided, detailing its size, appearance, texture, borders, and coloration. This information assists the detection system in making accurate assessments of the lesion's characteristics and potential risk factors for skin cancer. Furthermore, the text mentions the patient's concern about the possibility of skin cancer due to family history, highlighting the importance of early detection and monitoring for individuals with genetic predispositions to the disease. Overall, the text provides a concise

overview of the functionality, security features, and user guidelines for the skin cancer detection web application.

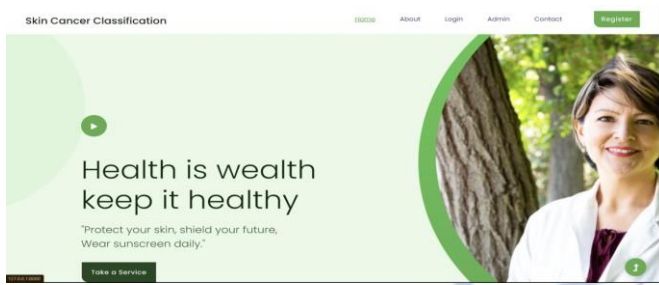


Figure 6. Home Page

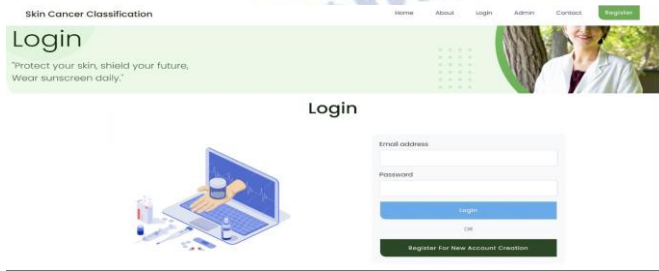


Figure 7. Login Page

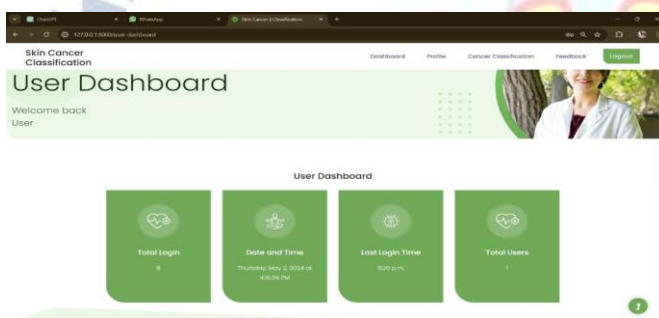


Figure 8. Dash Board

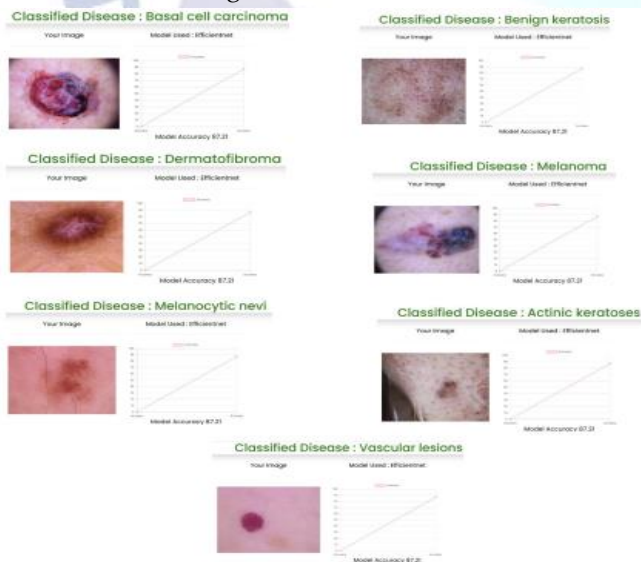


Figure 9. Cancer Classification

Model	Accuracy	F1 Score	Recall	Loss	Optimizer
Basic CNN	0.85	0.82	0.88	0.35	Adam
EfficientNet	0.92	0.89	0.94	0.20	Adam
Inception Model	0.91	0.88	0.93	0.22	Adam

Aspect	Basic CNN	EfficientNet	Inception Model
Architecture Complexity	Simple	More complex	Highly complex
Parameter Efficiency	Fewer parameters	Efficient scaling	More parameters
Performance	Moderate	State-of-the-art	Competitive
Training Speed	Fast	Efficient	Slower
Deployment	Easy	Requires resources	Challenging
Versatility	Versatile	Highly versatile	Versatile

Figure 10. Comparison Table

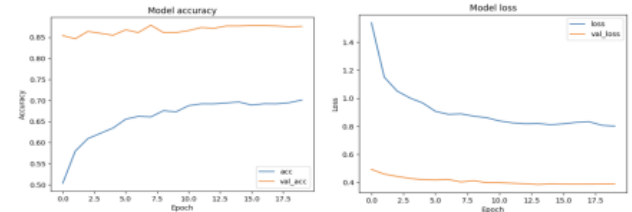


Figure 11. Model Accuracy and Loss Graphs

### CONFUSION MATRIX

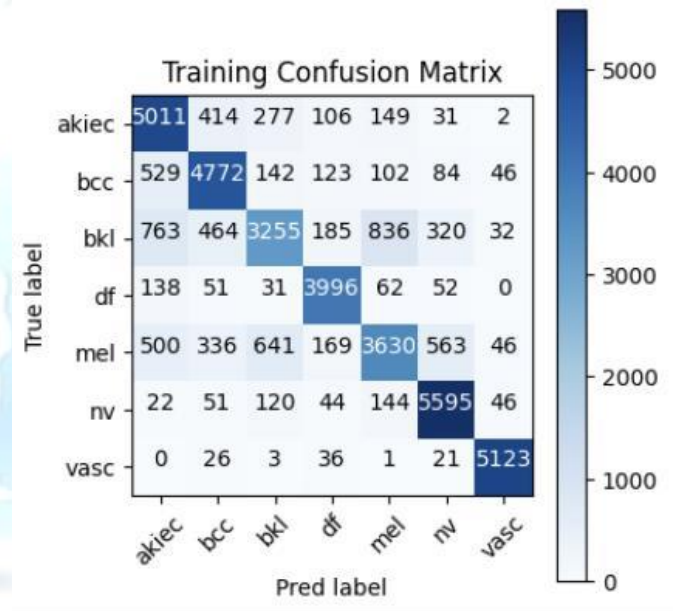


Figure 12. Efficient Net Confusion Matrix

A confusion matrix is essential for assessing machine learning classification models. It concisely summarizes how well the model's predictions match test data's true values. Comparison of anticipated and actual values in the confusion matrix shows model performance across classes or categories. Typical confusion matrix components:

1. True Positive (TP): The model successfully predicted a positive outcome, and the true value is positive. Thus, the model discovered positive class occurrences.
2. True Negative (TN): The model successfully predicted a negative outcome, and the true value is negative. The model correctly identified negatives.
3. False Positive (FP) (Type I Error): The model anticipated a positive outcome but got a negative one.



This is a Type I Error or false alarm. The model misclassified cases as affirmative. 4. False Negative (FN) (Type II Error): The model expected a negative result but got a positive one. This is a Type II Error, or missed detection. Positive class occurrences were missed by the model. Examining each confusion matrix cell's values can reveal the model's performance across classes. The confusion matrix may be used to assess the model's classification performance using accuracy, precision, recall, and F1 score. The confusion matrix helps evaluate a classification model's strengths and faults and enhance its performance.

## 7. FUTURE SCOPE AND CONCLUSION

### CONCLUSION

In conclusion, skin cancer is a worldwide health issue that requires early and precise detection to enhance patient outcomes. Traditional clinical evaluation methods face issues including long wait times and subjectivity, but deep learning (DL) can help. This work sought to enhance skin cancer categorization using DL prediction models by addressing class imbalance and improving interpretability for better decision-making. A smart healthcare system using optimal convolutional neural networks (CNNs) trained on the HAM10000 dataset to diagnose seven skin cancers was suggested via an Android app. DL algorithms are used to change skin cancer diagnosis by speeding up and improving diagnostics. The DL models can analyze massive volumes of data and find tiny patterns, improving diagnostic accuracy and reducing clinical evaluation subjectivity. The suggested online application for DL-powered skin cancer categorization has several benefits. First, it allows speedier diagnosis, which may reduce waiting times and improve patient outcomes. Second, DL algorithms evaluate data thoroughly and find patterns that humans may overlook, improving accuracy. Thirdly, objective studies based on trained models diminish diagnostic bias. Finally, openness into the model's decision-making process helps clinicians make educated patient care decisions. The promised benefits should be weighed against various drawbacks. Training data quality and diversity affect model accuracy, and data biases can skew findings. DL models' complicated decision-making processes are difficult to grasp, and data security and privacy regulations must be addressed. Internet connectivity and device availability

may also restrict web application accessibility. The web program should help doctors make clinical decisions, not replace them.

### FUTURE SCOPE

Expanding the skin cancer diagnosis system to include more kinds would improve patient care. Integrating the online application with EHR systems allows thorough assessments by combining important clinical data, and language accessibility provides inclusiveness and user-friendliness across varied populations. A smartphone app improves accessibility and encourages proactive skin health monitoring. The algorithm stays accurate and successful by learning and refining with new datasets. Deep learning-assisted skin cancer diagnosis might transform healthcare by promoting cooperation, responsible implementation, and continual development. It could enhance detection, treatment, and overall health.

### Conflict of interest statement

Authors declare that they do not have any conflict of interest.

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