



# Deep Learning Based Acne Vulgaris Detection And Personalized Skincare Recommendation

<sup>1</sup>Mr.D.Rajesh, <sup>2</sup>Saahitya K, <sup>3</sup>Asma A, <sup>4</sup>Devadharshini K

<sup>1</sup>Assistant Professor, <sup>2</sup>Student, <sup>3</sup>Student, <sup>4</sup>Student

<sup>1</sup>Computer Science and Engineering,

<sup>1</sup>Sri Manakula Vinayagar Engineering College,Puducherry,India

**Abstract:** Acne vulgaris significantly impacts individuals physical and emotional well-being, requiring timely diagnosis and grading for effective management. This research introduces an advanced acne detection and grading system powered by the YOLO V11 algorithm, renowned for accurate, real-time object detection. The system identifies and grades acne lesions such as blackheads, whiteheads, papules, pustules, and cysts while providing personalized skincare recommendations based on severity. It also offers geolocation-based dermatologist suggestions for cases needing professional care. By combining AI-driven analysis with human expertise, the system ensures accurate, scalable, and accessible acne management, enhancing outcomes and improving quality of life.

**Index Terms** - Acne vulgaris, YOLO V11, skin condition, accuracy.

## 1. INTRODUCTION

Acne vulgaris is a prevalent skin condition marked by the appearance of pimples, blackheads, whiteheads, and sometimes painful cysts. It often affects regions with a high density of sebaceous glands, including the face, chest, back, and shoulders. The condition arises when hair follicles become clogged with excess oil and dead skin cells, creating an environment that favors bacterial growth. Hormonal changes, particularly during puberty, frequently trigger these blockages, allowing *Propionibacterium acnes* bacteria to proliferate, which leads to inflammation and pus formation. This inflammatory response results in the red, swollen pimples characteristic of acne. In addition to hormonal factors, genetics, stress, diet, and the use of certain skincare products can also influence acne's severity and persistence. Though acne is not generally a serious medical threat, it can significantly impact an individual's self-esteem and, if left untreated, may cause permanent scarring. Treatment options vary based on severity and may include topical creams, oral medications, and lifestyle changes aimed at managing oil production and inflammation. In severe cases, dermatological procedures, such as laser therapy and chemical peels, are considered to help reduce scarring and improve skin texture.

## 2. OBJECTIVES

- **Develop an AI-Based System for Acne Detection:**

Leverage the YOLO V10 algorithm to detect and classify acne lesions in real-time, ensuring accurate identification of skin conditions. This will enhance diagnostic precision, allowing users to receive immediate, reliable feedback. The system aims to make acne detection faster, more accurate, and accessible.

- **Enable Severity Grading of Acne:**

Integrate a severity grading system to classify acne based on type and intensity, providing a detailed understanding of the condition. This feature supports informed decisions for appropriate treatment. By categorizing acne severity, users gain clarity on their skin's needs.

- **Provide Personalized Skincare Recommendations:**  
Analyse user-specific skin conditions to recommend tailored skincare solutions that address their unique acne concerns. These recommendations will improve the effectiveness of acne management. Personalized advice empowers users to optimize their skincare routine.
- **Create a User-Friendly Platform:**  
Develop an intuitive and accessible interface for seamless user interaction, making proactive skincare management easier. This approach reduces the physical and psychological burden of acne. The platform aims to enhance the overall quality of life for its users.

### 3. LITERATURE SURVEY

"Construction and Evaluation of a Deep Learning Model for Assessing Acne Vulgaris Using Clinical Images" by Yin Yang et al. (2023)

This study aims to enhance acne vulgaris diagnosis and treatment using a deep learning model aligned with the Chinese Guidelines for Acne Vulgaris Management. The facial images are divided into four regions and recombined to provide a comprehensive view of acne lesions. The Inception-v3 network classifies acne types like papules, pustules, nodules, and cysts and aligns its outputs with established treatment protocols. The model utilizes transfer learning through the pretrained ImageNet dataset to boost classification accuracy and recommends personalized treatment strategies. By improving diagnostic precision and treatment standardization, this model advances overall care outcomes for acne vulgaris patients. [1]

"Automatic Acne Severity Grading with a Small and Imbalanced Data Set of Low Resolution Images" by Rémi Bernhard et al. (2022)

Developing automated grading systems for acne vulgaris often requires large, high-resolution images from a diverse patient population. However, such datasets are costly and labor-intensive to obtain. To address this, this study proposes a deep learning model that assesses acne severity using low-resolution images and a smaller, imbalanced dataset with minimal labeling. Despite these challenges, the model achieved a 66.67% accuracy on the test set, demonstrating that it can perform comparably to more resource-intensive methods, offering a viable alternative for resource-constrained scenarios. [2]

"A Cell Phone App for Facial Acne Severity Assessment" by Jiaojia Wang et al. (2022)

This paper presents a system for automated acne severity assessment using smartphone applications. The model, developed with the Pyramid Scene Parsing Network (PSPNet), segments facial skin into regions to analyze location-specific acne. Acne-RegNet, a classification model, is then employed to identify and classify acne lesions within each region. The approach offers an accessible and detailed assessment of acne severity, empowering individuals to assess their own acne conveniently and promoting more effective self-care. [3]

"Machine Learning and Deep Learning Approach for Medical Image Analysis: Diagnosis to Detection" by Meghavi Rana and Megha Bhushan (2022)

This paper reviews machine learning (ML) and deep learning (DL) techniques used in medical image analysis, from diagnosis to detection. It discusses various algorithms and methodologies, such as CNNs, transfer learning, and generative models, for detecting diseases, segmenting images, and classifying abnormalities. The study also explores the challenges faced by these techniques, such as data quality, model interpretability, and computational demands. It highlights how ML and DL can transform healthcare by enabling efficient, accurate, and scalable diagnostic tools. [4]

"ScarNet: Development and Validation of a Novel Deep CNN Model for Acne Scar Classification with a New Dataset" by Masum Shah Junayed et al. (2021)

This study introduces ScarNet, a deep convolutional neural network (CNN) model designed for acne scar classification, which is built using a new, unique dataset. ScarNet provides high-precision classification of different types of acne scars and has been rigorously validated in clinical applications. By enhancing diagnostic accuracy and efficiency, ScarNet reduces the time and effort involved in acne scar classification tasks, making it a valuable tool for dermatological diagnosis and patient care. [5]

#### 4.PROBLEM IDENTIFICATION

- The algorithm Inception-v3 may struggle to generalize across diverse skin tones and variations in facial images, requiring high-quality, well-labeled clinical datasets, and possibly having limited adaptability for real-world clinical conditions.
- The algorithm's performance is limited by the use of low-resolution and imbalanced datasets, leading to lower accuracy, especially for less-represented acne severity grades.
- The algorithms Pyramid Scene Parsing Network (PSPNet), Acne-RegNet may encounter challenges in real-world scenarios due to inconsistent lighting, varied camera qualities, and different facial orientations, reducing classification accuracy in user-generated images.
- The deep CNN model faces limitations in generalizing to diverse datasets and acne scar types, especially underrepresented ones, reducing its effectiveness in real-world applications.

#### 5.EXISTING SYSTEM

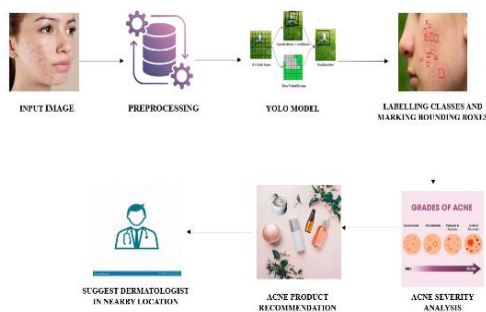
Acne scarring is a common consequence of acne vulgaris, affecting approximately 95% of individuals who experience acne. These scars result from the body's healing response, where collagen is either lost or produced in excess, leading to uneven skin texture and visible marks. Effective treatment relies on accurate classification of scar types, as each type may require a unique therapeutic approach. However, current scar classification practices are mostly manual and based on dermatologists' visual assessments. This process can be labour-intensive and is susceptible to variability between different observers, which can impact treatment consistency and effectiveness. To address these challenges, this study introduces an automated acne scar classification system using a deep learning model, named ScarNet. ScarNet leverages the power of Convolutional Neural Networks (CNNs) to analyze and classify acne scars with high accuracy. The dataset for this system includes 250 images of five distinct scar types, carefully labelled by four experienced dermatologists to ensure reliability. Prior to feeding the images into the CNN, pre processing steps are applied to enhance the quality and relevance of the data, which helps ScarNet focus on key scar features. The CNN architecture is optimized by tuning parameters such as the optimizer, loss function, activation functions, and batch size, aiming to maximize classification accuracy while keeping computational demands low. The performance of ScarNet is impressive, achieving an accuracy rate of 92.53%, a specificity of 95.38%, and a kappa score of 76.7%, which highlights the model's reliability and robustness. These metrics demonstrate that ScarNet can classify acne scars with high precision, reducing the potential for human error in diagnosis and ensuring a more consistent evaluation process. By automating scar classification, this system not only saves time for dermatologists but also improves the timeliness and precision of treatment decisions, ultimately enhancing patient outcomes in managing acne scars.

#### 6.PROPOSED SYSTEM

In the proposed system, acne prediction is achieved through the implementation of the YOLO V11 algorithm, a cutting-edge deep learning model specifically designed for real-time object detection. Upon receiving user-submitted images, the system analyzes the images to accurately identify and classify various types of acne lesions based on their characteristics and severity. Once the acne is detected, the system goes a step further by providing personalized skincare product recommendations tailored to the user's specific acne conditions. These suggestions are generated based on a comprehensive analysis of the detected lesions, ensuring that users receive effective solutions suited to their skin type. Additionally, the system facilitates user access to healthcare by recommending nearby dermatologists based on the user's geographical location, enabling them to seek professional advice and treatment if necessary. This dual functionality not only enhances the accuracy of acne detection but also supports users in managing their skin health more effectively by offering both immediate solutions and professional care options.

## 7. ARCHITECTURE OF PROPOSED SYSTEM

The architecture of the proposed system for acne detection and management consists of several interconnected modules, designed to streamline the process from image input to actionable recommendations. The workflow integrates advanced deep learning models with user-centric features to provide accurate, real-time insights into acne diagnosis and treatment. The key stages of the system architecture are as follows:



### ➤ Input Image

The process begins with capturing or uploading a facial image of the user. High-resolution images are recommended to ensure clear visualization of skin details and acne lesions. This image forms the basis for subsequent analysis.

### ➤ Preprocessing

Input images undergo preprocessing steps to enhance their quality and standardize them for analysis. Techniques such as noise reduction, contrast adjustment, and resizing are applied to ensure compatibility with the YOLO model. These steps improve the accuracy of lesion detection and severity grading.

### ➤ YOLO Model

The YOLO (You Only Look Once) algorithm serves as the core of the system, performing real-time object detection. It identifies acne lesions by marking bounding boxes and assigning appropriate labels to classify them into specific types, such as blackheads, whiteheads, or nodules. The model is trained to distinguish acne from normal skin features.

### ➤ Labeling Classes and Marking Bounding Boxes

After detection, acne lesions are labelled based on their classification, and bounding boxes are marked around the identified regions. This step provides a clear visual representation of affected areas, enabling users to understand the extent and type of their acne condition.

### ➤ Acne Severity Analysis

The identified acne lesions are analysed to assess their severity, ranging from mild to severe. The system calculates the severity grade based on factors such as lesion size, type, and distribution. This analysis helps users and dermatologists determine the appropriate course of action.

### ➤ Acne Product Recommendation

Based on the severity and type of acne, the system provides personalized skincare product recommendations. The suggested products are tailored to the user's needs, promoting effective acne management and improving skin health.

### ➤ Dermatologist Suggestion

For cases of severe acne, the system suggests consulting a dermatologist. Users are provided with information about dermatology clinics in their vicinity to facilitate timely professional intervention.

## 8. CONCLUSION AND FUTURE WORKS

In conclusion, the proposed acne detection and management system leverages the YOLO V11 algorithm to deliver precise and real-time acne identification, empowering users with actionable insights into their skin health. By accurately classifying acne types and severities, the system ensures that users receive skincare product recommendations tailored to their specific conditions, enhancing the effectiveness of suggested solutions. The added feature of locating nearby dermatologists based on user location further bridges the gap between technology and professional healthcare, offering users both immediate guidance and access to expert care. This comprehensive approach not only improves the user experience but also provides a holistic solution for managing acne, enabling users to make informed decisions and achieve better skin health outcomes.



Future work could focus on expanding the system's capabilities by incorporating additional skin conditions beyond acne, refining the model for higher accuracy across diverse skin tones, and integrating real-time user feedback to improve recommendation precision. Additionally, incorporating telemedicine features for virtual consultations could enhance accessibility to dermatologists.

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