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DEEP LEARNING-BASED PLANTAR PRESSURE ANALYSIS FOR EARLY DIABETIC FOOT RISK ASSESSMENT: A PROPOSED INTELLIGENT HEALTHCARE FRAMEWORK

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ABSTRACT :

Diabetes mellitus is a chronic health condition that can be associated with several long-term complications, including foot-related abnormalities and increased risk of diabetic foot complications. Early identification of abnormal plantar pressure patterns can provide useful information for preventive assessment and monitoring. Conventional assessment methods may depend on clinical examination and specialized equipment, motivating the development of intelligent computational approaches for automated plantar pressure analysis. This paper proposes a deep learning-based plantar pressure analysis framework for early diabetic foot risk assessment. The proposed framework consists of plantar pressure acquisition, preprocessing, foot-region segmentation, pressure-distribution analysis, deep feature extraction, risk classification, and explainable artificial intelligence. Convolutional neural networks and transfer-learning architectures such as ResNet and EfficientNet are proposed as candidate models for learning spatial characteristics from plantar pressure maps. An explainable AI module is incorporated to identify the plantar regions that contribute to the model's prediction. The proposed framework also includes an optional healthcare dashboard for visualizing pressure distributions and risk-assessment outputs. The methodology is designed as a conceptual and implementation-ready framework and does not claim experimental accuracy because a validated plantar-pressure dataset is not used in the present study. The work provides a foundation for future dataset-based validation and development of a non-invasive preliminary diabetic foot-risk screening system.

Keywords: *Plantar Pressure, Diabetic Foot, Deep Learning, Diabetes, Footprint Analysis, Risk Assessment, CNN, Explainable AI, Healthcare Artificial Intelligence.*

I. INTRODUCTION

Diabetes mellitus has become an important healthcare challenge because of its long-term effects on different organs and physiological systems. Foot-related complications represent an important area of concern because changes in foot loading and plantar pressure may be associated with increased risk of tissue damage and other complications. Early identification of abnormal plantar pressure patterns can support preventive monitoring and encourage timely clinical assessment.

Traditional foot assessment is generally dependent on clinical examination and specialized measurement systems. Although these approaches can provide valuable information, continuous and automated assessment can be difficult in environments where specialized equipment or clinical expertise is not readily available. Advances in artificial intelligence (AI), machine learning (ML), computer vision, and deep learning (DL) provide opportunities for automated analysis of complex spatial patterns.

The uploaded reference literature demonstrates how AI-based systems can accept visual input, perform automated analysis, classify the input, and provide an understandable output through an interactive interface. In the uploaded fake-product authentication study, users can provide an image through camera capture or image upload, after which an AI backend analyzes visual characteristics and produces an authenticity result. The study also integrates predictive analytics, dashboards, reporting, and accessibility-oriented functionality.

The same general principle can be adapted to healthcare applications. Instead of analyzing product images, the proposed system analyzes plantar pressure distributions and estimates diabetic foot risk.

The objective of this study is therefore to design a deep learning-based framework capable of processing plantar pressure information and generating an early risk assessment.

The proposed research does not attempt to replace medical diagnosis. Instead, it focuses on developing an intelligent preliminary screening and decision-support framework.

II. PROBLEM STATEMENT

Existing diabetic-foot assessment procedures may require clinical expertise and specialized measurement systems. There is a need for an intelligent framework that can automatically analyze plantar pressure distribution and identify patterns potentially associated with increased diabetic foot risk.

The problem addressed in this research is:

“How can deep learning be applied to plantar pressure distributions to develop an automated framework for early diabetic foot-risk assessment?”

The proposed framework aims to address this problem through automated preprocessing, spatial pressure analysis, deep learning, classification, and explainable prediction.

III. OBJECTIVES

The major objectives of the proposed study are:

- To design a plantar pressure data acquisition and preprocessing framework.
- To identify spatial pressure characteristics from plantar-foot measurements.
- To develop a deep learning architecture for plantar pressure analysis.
- To investigate multiple deep learning models for risk classification.
- To incorporate explainable AI into the proposed prediction framework.
- To design an understandable user interface for presenting risk information.
- To provide a foundation for future clinical and dataset-based validation.

IV. LITERATURE-BASED MOTIVATION

A. Artificial Intelligence-Based Analysis

AI has increasingly been incorporated into intelligent applications to automate analysis and support decision-making. The uploaded literature includes systems based on AI, machine learning, natural language processing, deep learning, and predictive analytics.

The uploaded smart resume analysis paper discusses the use of machine learning, deep learning, transformer architectures, and explainable AI for automated analysis.

Although the application domain is different, the methodological concept of automatically extracting meaningful representations from complex input is applicable to healthcare.

B. AI-Based Image Analysis

The uploaded authentication study demonstrates an image-analysis workflow in which users capture or upload an image and the backend performs AI-based analysis.

This provides a useful conceptual basis for the proposed system:

Input → preprocessing → AI analysis → classification → result

The proposed work replaces product authentication with plantar pressure analysis.

C. Real-Time Healthcare-Oriented Interface

The uploaded reference also demonstrates the integration of automated analysis with dashboard visualization and real-time information management.

A similar interface can be adapted to display plantar pressure maps, risk predictions, and explanatory visualizations.

V. RESEARCH GAP

The uploaded reference papers address areas including:

- AI-based analysis,
- predictive analytics,
- image processing,
- intelligent dashboards,
- accessibility,
- blockchain-based verification,
- natural language processing,

- transformer-based models,
- embedded hardware.

However, the supplied files do not present a complete framework specifically focused on deep learning-based plantar pressure analysis for early diabetic foot-risk assessment.

Therefore, the proposed study introduces a healthcare-specific framework that combines:

$$\text{Plantar Pressure Analysis} + \text{Deep Learning} + \text{Risk Assessment} + \text{Explainable AI}$$

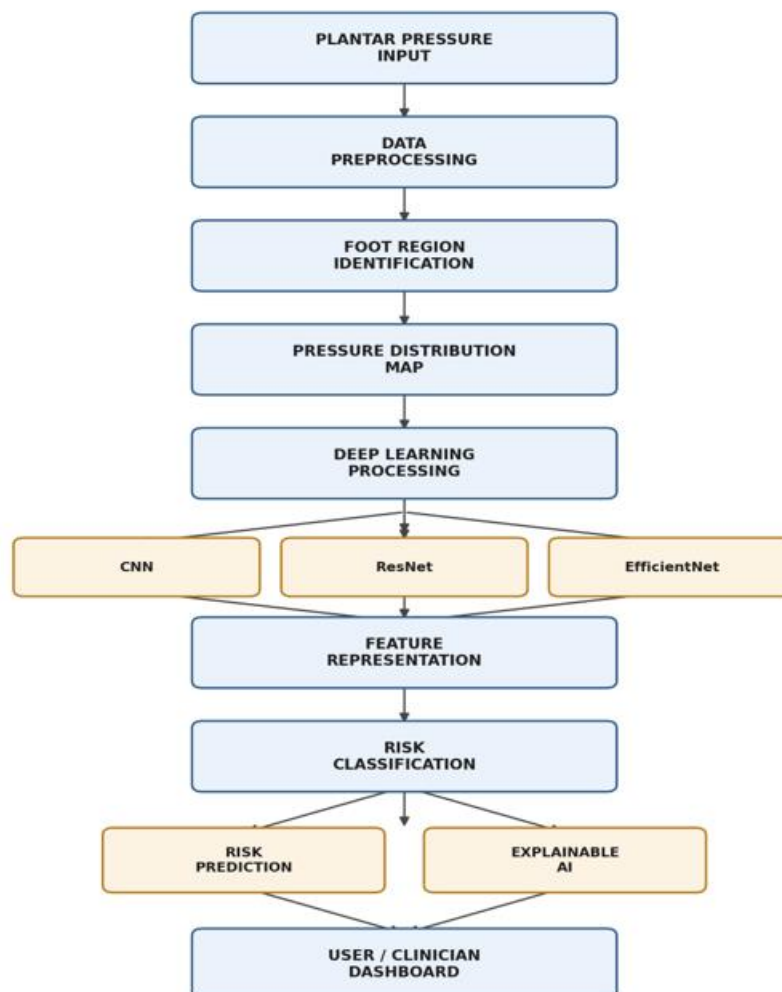
The primary contribution is the design of an integrated methodology rather than the presentation of experimentally measured accuracy.

VI. PROPOSED SYSTEM

The proposed system consists of the following major stages:

- Plantar pressure acquisition
- Data preprocessing
- Foot-region identification
- Pressure distribution analysis
- Deep learning feature extraction
- Risk classification
- Explainable AI
- Healthcare dashboard

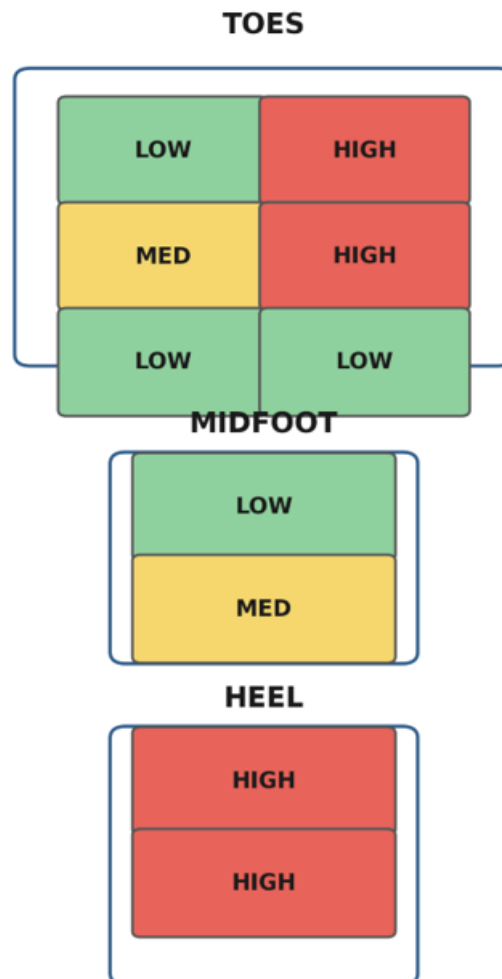
Overall workflow:



VII. PLANTAR PRESSURE DATA ACQUISITION

The proposed framework requires plantar pressure measurements obtained from an appropriate pressure-sensing system.

The input may be represented as a two-dimensional pressure map in which each pixel or sensing element corresponds to a pressure value. A conceptual representation is:



The actual pressure values and sensor configuration depend on the hardware used in future implementation.

VIII. DATA PREPROCESSING

Before deep learning analysis, the raw pressure measurements should undergo preprocessing. The proposed preprocessing pipeline includes:

8.1 Noise Removal

Unwanted sensor noise should be reduced to improve the quality of the pressure map.

8.2 Normalization

Pressure values should be normalized to an appropriate numerical range.

8.3 Foot Detection

The relevant plantar region should be separated from the background.

8.4 Resizing

Pressure maps should be converted to a consistent input size required by the selected deep learning architecture.

8.5 Left and Right Foot Separation

If both feet are recorded simultaneously, they should be separated to enable independent and comparative analysis.

IX. PLANTAR REGION ANALYSIS

The plantar surface can be divided into meaningful regions for computational analysis. The proposed framework considers:

- toe region,
- forefoot region,
- midfoot region,
- heel region.

Pressure-related characteristics can then be extracted from each region. Potential computational characteristics include:

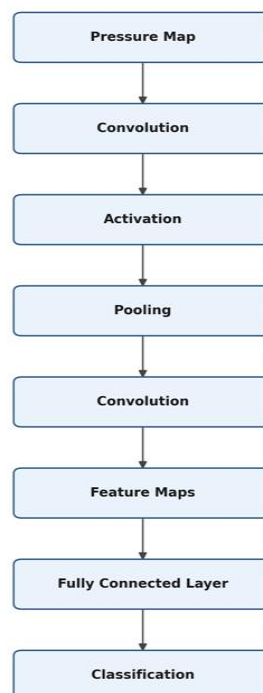
- maximum pressure,
- mean pressure,
- pressure variation,
- pressure concentration,
- contact area,
- regional pressure distribution,
- left-right pressure difference.

These characteristics can be used as complementary information to deep learning features.

X. DEEP LEARNING MODEL

10.1 Convolutional Neural Network

A CNN can be used as the baseline architecture. The general CNN process is:



The CNN learns spatial characteristics directly from the plantar pressure map.

10.2 ResNet

ResNet can be investigated to determine whether deeper feature extraction improves plantar-pressure representation. Residual connections can help construct deeper networks while reducing degradation problems associated with conventional deep architectures.

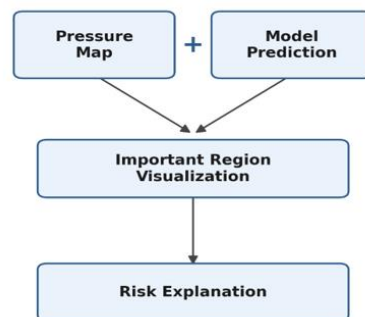
10.3 EfficientNet

EfficientNet can be evaluated as a computationally efficient alternative. Its use may be particularly appropriate if the eventual objective is deployment on resource-constrained healthcare devices.

XI. EXPLAINABLE ARTIFICIAL INTELLIGENCE

An important component of the proposed framework is explainability. Healthcare users may require more information than a simple “High Risk” prediction.

Therefore, an explainability mechanism such as Grad-CAM can be investigated. The proposed output is:



For example, the interface could highlight pressure regions that contributed strongly to the model prediction. This approach is consistent with the importance of transparency and explainability discussed in the uploaded AI literature.

The explanation should be considered a model interpretation rather than a clinical diagnosis.

XII. PROPOSED RISK CLASSIFICATION

Instead of claiming that the system directly diagnoses diabetes, the proposed framework uses a risk-assessment formulation. A conceptual classification can be:

Class	Description
Low Risk	Pressure distribution does not indicate the target high-risk pattern
Moderate Risk	Some potentially abnormal pressure characteristics are observed
High Risk	Pressure distribution demonstrates patterns associated with increased risk

The actual class definitions must ultimately be established using an appropriately labeled clinical dataset.

XIII. USER INTERFACE

The proposed system can provide a simple dashboard.

Dashboard components

- Patient/record identifier
- Plantar pressure map
- Left-foot pressure distribution
- Right-foot pressure distribution
- Regional pressure information
- Predicted risk category
- Explainable pressure visualization
- Historical records

The uploaded reference demonstrates that AI systems can combine automated processing with dashboard-based visualization and real-time data presentation.

XIV. PROPOSED ALGORITHM

Algorithm: Plantar Pressure Risk Assessment

Input: Plantar pressure measurement

Output: Risk category and explanatory visualization

- Step 1: Acquire plantar pressure data.
- Step 2: Validate the input measurement.
- Step 3: Remove noise and normalize pressure values.
- Step 4: Detect the plantar foot region.
- Step 5: Divide the plantar surface into predefined regions.
- Step 6: Generate the normalized pressure map.
- Step 7: Pass the pressure map to the selected deep learning model.
- Step 8: Extract spatial deep features.
- Step 9: Perform risk classification.
- Step 10: Generate an explainability map.
- Step 11: Display the predicted risk and relevant pressure regions.
- Step 12: Store the result securely for future monitoring.

XV. SYSTEM SECURITY AND PRIVACY

Because the proposed system concerns healthcare information, privacy should be considered during implementation.

The system should avoid unnecessary personal information and use an anonymized record identifier. Potential security mechanisms include:

- authenticated user access,
- encrypted data transmission,
- encrypted storage,
- access control,
- anonymized records,
- audit logging.

The uploaded authentication paper demonstrates the broader value of secure and traceable data management in intelligent systems. It uses blockchain-backed records to provide integrity and traceability.

Blockchain is not required for the proposed plantar-pressure model and should only be added if there is a specific research requirement for immutable medical-record auditing.

XVI. EXPECTED CONTRIBUTIONS

Since this is a proposed-framework paper and no experimental dataset is used, the contributions are methodological rather than numerical. The proposed research contributes:

- A deep learning-oriented framework for plantar pressure analysis.
- A structured preprocessing methodology.
- Regional plantar-pressure analysis.
- A risk-assessment formulation rather than unsupported diagnostic claims.
- Integration of explainable AI.
- A conceptual healthcare dashboard.
- A foundation for future clinical dataset validation.

XVII. DISCUSSION

The proposed framework demonstrates how deep learning concepts used in intelligent systems can be transferred to plantar-pressure healthcare analysis.

The uploaded authentication study illustrates an architecture in which an input is captured, analyzed automatically, classified, and presented to users. The proposed research applies a similar computational workflow to plantar pressure.

However, healthcare applications require additional considerations. A model designed for diabetic foot-risk assessment must ultimately be evaluated against clinically meaningful labels. Therefore, the proposed framework deliberately avoids reporting unsupported performance values.

The absence of a dataset in the current study is a limitation, but it does not prevent the development of a conceptual research framework. Future experimental work can implement the proposed architecture and evaluate it using an appropriately collected or publicly available plantar-pressure dataset.

XVIII. LIMITATIONS

The major limitations of the present study are:

- No experimental plantar-pressure dataset is used.
- No numerical model-performance results are reported.
- Clinical validation has not been performed.
- The exact pressure-sensing hardware is not specified.
- The proposed risk categories require clinical labeling.
- The proposed framework should not be considered a diagnostic system.

XIX. FUTURE WORK

Future work will focus on:

19.1 Dataset Development

A sufficiently large plantar-pressure dataset should be collected with appropriate clinical labeling.

19.2 Model Training

CNN, ResNet, EfficientNet, and lightweight architectures can be trained and compared.

19.3 Multimodal Learning

Plantar pressure can be combined with:

- footprint images,
- demographic information,
- clinical measurements,
- gait information.

19.4 Temporal Analysis

Sequential pressure measurements can be analyzed to study plantar loading during walking.

19.5 Mobile Deployment

A lightweight model can be deployed on mobile or edge devices for accessible screening.

19.6 Clinical Validation

The system should eventually be evaluated with clinicians and an appropriately designed clinical study.

XX. CONCLUSION

This paper proposed a deep learning-based plantar pressure analysis framework for early diabetic foot-risk assessment. The framework consists of plantar pressure acquisition, preprocessing, regional analysis, deep learning feature extraction, risk classification, explainable AI, and healthcare-oriented visualization.

The proposed framework is motivated by the increasing use of AI-based automated analysis and the system-design principles demonstrated by the supplied reference literature. The uploaded intelligent authentication study demonstrates how automated AI analysis, real-time processing, reporting, visualization, and user-oriented interfaces can be integrated into a complete intelligent system.

In the proposed healthcare framework, plantar pressure replaces the original application input and becomes the basis for computational risk assessment. The framework emphasizes explainability and preliminary screening rather than claiming autonomous medical diagnosis.

Because no validated dataset is used in the present work, numerical accuracy and clinical performance are intentionally not reported. The proposed architecture provides a foundation for future implementation, dataset development, experimental evaluation, and clinical validation.

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