

MediMitra: Advanced Pill Identification and Guidance System Using Deep Learning

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Abstract - MediMitra is an innovative application that harnesses advanced deep learning techniques to provide users with a comprehensive healthcare assistance solution. Designed to facilitate pill identification, recommend home remedies based on symptoms, and locate nearby pharmacies, the system comprises three integral modules. The Pill Identifier allows users to upload pill images for rapid analysis, matching visual features like color, shape, and imprint against a vast database, offering detailed information on each pill, including its name, uses, and precautions. The Symptoms-Based Home Remedies Recommendation module enables users to input their symptoms, where a Consultation Algorithm evaluates this data and suggests tailored, non-invasive home remedies alongside lifestyle and dietary recommendations. Finally, the Map Navigator utilizes location-based services to help users easily find nearby pharmacies for prescribed medications. By integrating artificial intelligence with healthcare, MediMitra bridges the gap between immediate health needs and professional medical guidance, empowering users to manage their health proactively and effectively.

Key Words: Pill Identification, Symptom-Based Recommendation, Deep Learning, Pharmacy Locator, Home Remedies Recommendation.

1. INTRODUCTION

Medication errors are a serious issue that can have significant consequences for patient health and safety. With the vast variety of pharmaceutical products available, the accurate identification of pills has become a critical task for both healthcare professionals and patients. This task is made even more challenging by the fact that many pills share similar shapes, colors, and sizes, which makes manual identification difficult and prone to errors [1]. Medication errors can occur at multiple points, such as when a doctor prescribes the wrong drug, a pharmacist makes a dispensing error, or a patient takes medication incorrectly. Even with the best efforts of healthcare professionals, errors still happen. Problems may arise from mishandling medications, such as when an individual comes across an unknown pill or accidentally confuses two medications [2]. Issues like hard-to-read labels and damaged packaging can also contribute to mistakes. Moreover, self-medication, particularly when buying drugs without professional advice, increases the

likelihood of errors, especially in settings without a pharmacist to offer guidance.

When medication errors cause harm to a patient, they result in adverse drug events (ADEs). Importantly, ADEs associated with medication errors are preventable. Research has shown that these errors most frequently occur during the prescribing phase (56%) and the administration phase (34%) in hospital settings, where there is often no intermediary to detect or prevent the error before it reaches the patient. Factors such as the similarity in drug appearance or drug names, known as Look-Alike Sound-Alike (LASA) errors, pose a serious risk, particularly at the level of pharmacists or physicians. Such errors can put patient safety at significant risk, emphasizing the need for improved pill identification systems.

While traditional methods have laid the groundwork for pill identification, recent advancements in deep learning have opened up new possibilities for improving accuracy and performance. Deep learning algorithms, particularly convolutional neural networks (CNNs), are highly effective in feature extraction and object detection tasks, making them well-suited for robust pill recognition even in uncontrolled environments [3]. Various studies have demonstrated the potential of deep learning models to identify pills from images captured on mobile devices, offering a valuable tool for both patients and healthcare professionals. However, ensuring consistent performance across diverse conditions such as varying lighting, backgrounds, and image quality remains a significant challenge.

The design, implementation, and evaluation of MediMitra, a web-based application that leverages CNNs for advanced image processing. The system is designed to automatically identify pills based on their visual characteristics, aiming to mitigate the risk of medication errors and streamline the identification process. By using CNNs, MediMitra can accurately process and analyze pill images, extracting key features such as shape, color, and size to make precise identifications. This paper evaluates the performance of MediMitra, discusses the challenges faced in its development, and highlights its potential to significantly reduce medication errors and improve patient safety.

2. LITERATURE REVIEW

Supawadee Srikamdee et al. [1], developed a mobile-based application to automatically identify pills based on their shape and color using Mask-RCNN for shape recognition and K-means clustering for color detection. The system achieves high accuracy in pill identification under unconstrained real-world conditions.

Po-Tseng Wu et al. [2], propose a system to accurately recognize similar-looking round pills using an Attention-YOLO (AY) model. This model combines the YOLO v3 architecture with attention mechanisms to enhance feature extraction and spatial localization. Trained on 7,063 pill images across four shape types, the AY model achieves 92.28% accuracy. The system helps medical professionals accurately identify round pills, reducing medication errors and improving patient care.

Roy Chaoming Hsu and Po-Cheng Su [3], propose a system using YOLOv4 deep learning to reduce medication errors by identifying pills and verifying them against QR-encoded prescriptions. Running on Nvidia's Jetson Nano, it enables real-time processing. The system marks a shift from traditional image recognition to advanced techniques like CNNs and YOLO for efficient pill detection, improving patient safety.

Amir Reza Ashraf et al. [4], explore the use of code-free deep learning (CFDL) for developing a pill recognition model to reduce medication errors in clinical settings. Using Microsoft Azure's Custom Vision platform, the model was trained on images of the top 30 most dispensed pills from three hospitals, achieving high precision (98.7%) and recall (95.1%) in tests. However, performance varied on different platforms, especially offline mobile devices. The study concludes that CFDL is a feasible and cost-effective approach, though challenges remain in real-world clinical use, requiring further research to improve adaptability and image processing techniques.

Mohan Wang et al. [5], propose a novel model to accurately classify white pills, which are visually challenging to differentiate under visible light. By utilizing multi-band infrared (IR) imaging and enhancing the YOLOv5s model with multimodal fusion techniques, the study achieves higher detection accuracy. The model incorporates six IR channels and integrates an auxiliary detection branch and attention mechanisms, resulting in improved feature precision and a notable increase in classification performance.

Nilesh Kumar Maurya et al. [6], propose a system that helps users locate nearby medical stores based on their current location. The system utilizes GPS and mapping technologies to provide users with real-time information about the closest pharmacies and medical stores. The main objective is to

make it easier for individuals to find and access essential medical supplies, especially in emergencies.

Matiwos Tekalign Wondmagegn et al. [7], developed a system to help patients in developing countries find pharmacies with their prescribed medications. The framework uses GPS and Google Maps to address issues like distance, drug cost, travel time, and pharmacy hours. Tested in Mbarara, Uganda, it efficiently identifies nearby pharmacies with the required drugs, improving patient access and reducing medication delays.

Prajakta Khairnar et al. [8], presents a system that predicts diseases and suggests medications based on user-entered symptoms. Using Decision Tree, Random Forest, and Naive Bayes algorithms, the system found Naive Bayes to be the most accurate, with a 98.12% accuracy rate. Designed to assist users with limited access to healthcare, it provides quick and accurate medication recommendations for minor symptoms through an easy-to-use interface.

Benjamin Stark, Constanze Knahl et al. [9], explores various approaches to medicine recommendation systems in healthcare. It reviews existing systems, comparing ontology-based and machine learning-based solutions, which assist healthcare providers in selecting appropriate medications for patients. The paper highlights the need for scalable systems that consider dosage recommendations and emphasizes future research on integrating graph databases for better scalability and performance. The study concludes by identifying gaps and suggesting enhancements to improve the accuracy and usability of these systems.

3. PROPOSED SYSTEM

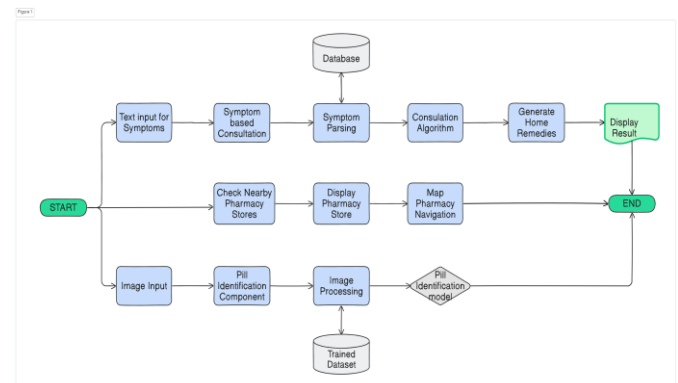


Fig -1: Proposed System Architecture

The MediMitra system offers a comprehensive healthcare solution combining pill identification and symptom-based consultation. The Pill Identification feature allows users to upload images of medication, which are processed using advanced AI models (CNN). These models are trained to recognize pill attributes such as shape, size, and color, ensuring accurate identification to verify prescriptions and prevent medication errors [5]. Alongside this, the Home

Remedies Module provides users with tailored advice based on their inputted symptoms. Using a symptom parsing algorithm, the system generates home remedies, precautionary measures, diet plans, and workout suggestions for mild conditions. All recommendations are sourced from a reliable medical database, offering safe and effective guidance. Moreover, the system integrates with nearby pharmacy services to help users obtain necessary over the counter medications [6]. MediMitra leverages AI and deep learning to deliver a user-friendly, real-time health management system that enhances patient safety and well-being.

The system architecture is depicted in Fig .1. where the process flow is divided into two main functionalities: symptom-based consultation and pill identification.

The system architecture is composed of five core modules: input, preprocessing, identification, symptom analysis, and recommendation. The input module captures pill images, and the preprocessing module ensures image quality is optimized. A convolutional neural network (CNN) in the identification module is used for pill classification. Additionally, the system analyzes user-reported symptoms to suggest appropriate home remedies. Finally, the recommendation module provides detailed guidance on home remedies and pharmacy locations where the medicine can be obtained [7].

3.1 WORKFLOW FOR PILL IDENTIFICATION MODULE

Step 1: The user captures an image of the pill and uploads it to the system.

Step 2: The system preprocesses the image by adjusting contrast, size, and sharpness for better feature extraction.

Step 3: A CNN model is employed to classify the pill based on visual features (color, shape, and imprint).

Step 4: If the user inputs symptoms, the system provides home remedies that align with those symptoms.

Step 5: The system offers pill identification results and relevant home remedies, including dosage information and details of nearby pharmacies.

3.2 WORKFLOW FOR SYMPTOMS-BASED HOME REMEDIES MODULE

Step 1: User inputs symptoms into the system.

Step 2: System processes symptoms using the Consultation Algorithm.

Step 3: Algorithm evaluates symptoms and classifies the condition as non-critical.

Step 4: System generates home remedy suggestions based on the symptoms.

Step 5: System provides additional lifestyle, diet, and workout recommendations.

4. METHODOLOGY

The MediMitra system is developed using a modular approach, integrating advanced deep learning techniques, symptom-based analysis, and geolocation services. The system is divided into three main modules: Pill Identification, Symptoms-Based Home Remedies Recommendation, and Nearby Pharmacy Locator. Each module works cohesively to provide a comprehensive healthcare solution.

4.1 IDENTIFICATION MODULE

Data Collection: A dataset of pill images is collected, consisting of various shapes, colors, and imprints from public medical repositories and custom sources.

Preprocessing: Images are preprocessed to ensure uniformity in size, resolution, and background, applying techniques such as resizing, noise reduction, and normalization.

Model Training: A Convolutional Neural Network (CNN) is employed for feature extraction and classification. The model is trained using the preprocessed pill images, where key features such as color, shape, and imprints are used for classification.

Identification Process: When a user inputs an image of a pill, the image is processed and classified using the trained CNN model, matching it with the existing pill database to provide information on the medication.

4.2 SYMPTOMS-BASED HOME REMEDIES RECOMMENDATION MODULE

Symptom Input: Users provide symptoms through text input or selection from a predefined list.

Consultation Algorithm: The algorithm parses the symptoms, categorizes the condition, and evaluates its severity. If the condition is non-critical, the system proceeds to suggest remedies.

Remedies and Suggestions Generation: Based on the evaluated symptoms, the system generates traditional home remedies, lifestyle adjustments, diet plans, and workout recommendations tailored to the user's condition. This is achieved through a curated database of remedies and best practices for common non-critical conditions.

Feedback Mechanism: Users may provide feedback on the remedies effectiveness, which can be used to refine future suggestions.

4.3 NEARBY PHARMACY LOCATOR MODULE

Location Tracking: The module uses the device's GPS or location services to identify the user's current location.

Pharmacy Database: The system accesses a database of pharmacies, including details like location, contact information, and medication availability.

Pharmacy Recommendations: Based on the user's location and the identified medication from the Pill Identification Module, the system recommends nearby pharmacies where the required medicine is available.

5. EXPERIMENTAL RESULT

MediMitra is designed to provide accurate results across all its modules, ensuring user satisfaction through its ease of use and effectiveness. The system aims to reduce the time spent on medication management tasks by, improving overall efficiency. It will likely lead to measurable health outcomes, such as fewer medication errors and better adherence to treatment plans. Designed for scalability, MediMitra can adapt to various healthcare settings and is projected to be cost-effective, providing savings for both providers and patients. Overall, these results highlight its potential to significantly enhance healthcare experiences.

6. CONCLUSION

This paper presents, MediMitra, an integrated healthcare solution that combines a deep learning-based pill identification system, a personalized home remedies recommendation system, and a real-time healthcare facility navigation tool. The pill identification module employs a Convolutional Neural Network (CNN) developed using Keras and TensorFlow, trained on a large dataset of labeled pill images. This dataset underwent pre-processing to normalize input data and enhance image features such as color, shape, and imprint. The home remedies recommendation system aligns user-reported symptoms with appropriate remedies, while the map navigator offers GPS-based guidance to nearby healthcare facilities. By addressing medication safety, self-care, and healthcare accessibility, MediMitra provides an integrative approach to health management. Future enhancements will focus on expanding the datasets, and integrating the system into broader healthcare platforms to maximize its utility and impact.

7. REFERENCES

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