

AI-Based Intelligent Medical Diagnosis Assistance System Using Voice Processing and Gemini API

Nisha, Payal¹, Niharika Rai², Mr. Devendra Sharma³

^{1,2} *Department of Artificial Intelligence and Data Science IIMT College of Engineering*

³ *Guide, Department of Artificial Intelligence and Data Science IIMT College of Engineering*

Abstract—The integration of Artificial Intelligence in healthcare has significantly improved diagnostic processes and patient interaction systems. This paper presents an AI-based intelligent medical diagnosis assistance system that utilizes voice input, natural language processing, and machine intelligence to assist users in identifying possible medical conditions. The system converts speech into text, summarizes patient symptoms, predicts diagnoses, and recommends medicines. The Google Gemini API is used for AI processing, while Selenium-based web scraping is used to build a medicine dataset. The system aims to provide an accessible, efficient, and semi-automated healthcare support solution.

By integrating natural language understanding and machine learning capabilities, the system ensures accurate interpretation of user inputs while enhancing accessibility for individuals with limited technical or medical knowledge. The proposed solution aims to reduce diagnostic delays, improve early detection of diseases, and support healthcare professionals by offering data-driven insights. Additionally, it promotes inclusivity by enabling hands-free interaction, making healthcare guidance more accessible in remote and underserved areas.

This intelligent framework demonstrates the potential of combining voice technology with advanced AI models to create scalable, efficient, and patient-centric healthcare solutions.

I. INTRODUCTION

Healthcare systems often require efficient tools to assist patients in early-stage diagnosis. Many users struggle to accurately describe symptoms, which leads to delays in diagnosis. With advancements in AI and Natural Language Processing (NLP), automated systems can now understand human language and provide intelligent responses.

This project introduces a voice-enabled AI system that simplifies symptom input through speech. The system processes the input using advanced AI models and provides meaningful insights such as symptom summaries, possible diagnoses, and medicine recommendations.

The integration of Artificial Intelligence (AI) in healthcare is reshaping the way medical services are delivered, enabling faster, more accurate, and accessible diagnostic support. With the growing demand for early disease detection and the increasing burden on healthcare systems, there is a need for intelligent solutions that can assist both patients and medical professionals. This project introduces an AI-based medical diagnosis assistance system that utilizes voice

processing and the advanced capabilities of the Gemini API to provide realtime preliminary health analysis.

II. RELATED WORK

Previous research in AI-based diagnosis systems has focused on rule-based systems and machine learning models.

However, these systems often lack contextual understanding.

Recent advancements in large language models have improved contextual reasoning, enabling better diagnosis prediction. This project integrates modern AI models with realworld datasets to improve performance.

Recent advancements in Artificial Intelligence (AI) and healthcare technologies have led to the development of intelligent diagnostic systems aimed at improving early disease detection and patient support. Prior research has explored symptom-based

diagnosis models using machine learning algorithms and rule-based expert systems to provide preliminary medical insights. These systems primarily rely on structured inputs such as text or predefined symptom checklists. .

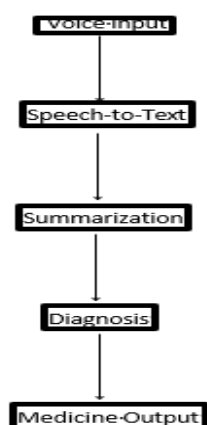
Recent studies have also leveraged advanced AI platforms such as the Gemini API to improve the accuracy and contextual understanding of medical queries. This project builds upon existing research by combining voice processing with advanced AI capabilities to create a more efficient, accessible, and intelligent medical diagnosis assistance system.

III. METHODOLOGY

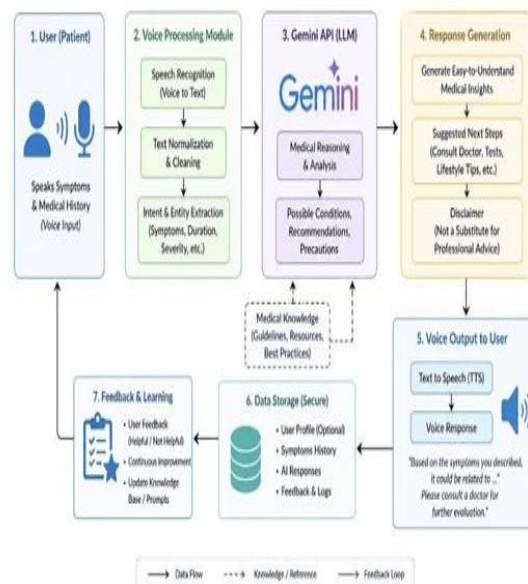
The proposed system follows a structured and modular approach to deliver accurate and efficient preliminary medical assistance. The methodology begins with capturing user input through voice, which is processed using speech recognition techniques to convert spoken symptoms into textual data. The converted text is then preprocessed using techniques from Natural Language Processing, including tokenization, noise removal, and intent recognition, to extract meaningful medical information. The refined input is passed to the Gemini API, which analyzes the symptoms using advanced AI models and generates possible diagnoses along with relevant suggestions.

The methodology also incorporates validation mechanisms to ensure that the extracted symptoms are relevant and meaningful. Ambiguous inputs are refined using contextual analysis to improve the quality of predictions.

A. Flowchart



AI-Based Intelligent Medical Diagnosis Assistance System
Using Voice Processing and Gemini API



IV. DATASET DESCRIPTION

The dataset used in this project was collected through web scraping from online medical platforms. Selenium was used to automate data extraction, which included medicine names, usage instructions, disease associations, and dosage information.

The dataset contains approximately 500+ medicine entries and was preprocessed to remove duplicates and inconsistencies. This dataset plays a critical role in ensuring realistic and accurate recommendations.

The effectiveness of the proposed system relies on a wellstructured and diverse medical dataset that supports accurate symptom analysis and preliminary diagnosis.

The data includes structured attributes such as symptom names, associated conditions, severity levels, and possible recommendations. To enhance the system's understanding, unstructured textual data is also incorporated and processed using techniques from Natural Language Processing. This enables the model to interpret natural language inputs obtained through voice processing.

Additionally, the system leverages the capabilities of

the Gemini API, which utilizes large-scale pre-trained data to improve contextual understanding and generate meaningful outputs.

V. ALGORITHM

The proposed system follows a systematic algorithm to process user input and generate intelligent diagnostic suggestions. Initially, the system captures voice input from the user and converts it into text using speech recognition

VI. IMPLEMENTATION

The system is implemented using React for the frontend and a backend server that integrates the Gemini API. Prompt engineering is used to guide AI outputs. The system ensures fast response time and structured outputs.

The implementation of the proposed system is carried out using a modular and scalable architecture that integrates voice processing, data handling, and AI-driven analysis. The frontend interface is designed to provide a simple and intuitive user experience, allowing users to input symptoms through voice commands. Speech input is captured and converted into text using speech recognition libraries.

The processed text is then handled using techniques from Natural Language Processing to extract relevant medical information. The backend system is developed using

programming languages such as Python, where the cleaned input is passed to the Gemini API for intelligent analysis and response generation techniques. The extracted text is then preprocessed using methods from Natural Language Processing, including tokenization, stop-word removal, and keyword extraction to identify relevant symptoms.

Next, the processed input is analyzed to map the identified symptoms with the underlying dataset. The refined data is then passed to the Gemini API, which applies advanced AI models to interpret the context and generate potential diagnoses along with recommendations. Finally, the system ranks the possible outcomes based on relevance and confidence levels, and displays the results in a user-friendly format. Basic validation and filtering mechanisms are incorporated to enhance accuracy and reduce ambiguity.

Algorithm Steps:

- 1) Capture voice input
- 2) Convert speech to text
- 3) Generate summary using AI
- 4) Send summary to Gemini API
- 5) Predict diagnosis
- 6) Map diagnosis to medicines
- 7) Display results

A structured dataset is integrated to support symptom mapping and validation, ensuring more reliable outputs. The system also includes error handling and basic validation mechanisms to manage incomplete or ambiguous inputs. The final output is displayed in a clear and user-friendly format, providing possible diagnoses and suggestions.

Overall, the implementation focuses on creating a responsive, efficient, and accessible solution by combining voice technology with advanced AI capabilities in a real-time environment.

Error handling mechanisms are integrated to manage incomplete or unclear inputs. The system prompts users for clarification when necessary, improving overall interaction quality.

The use of modern web technologies ensures compatibility across multiple devices, including desktops and mobile platforms.

Mr Kush Rajput
Current Visit

Age: 28 yrs • Male

Next Patient v

K

No vitals recorded today
 + Add Vitals
+ Smart Plans v

Recording 1

English (India)

Record

Clear

Waveform will appear here...

Notes

Record note

Auto

Enter consultation notes, examination findings, patient observations...

Consultation Question Guide

Transcript History 1

AI Consultation Summary (Generated)

English (India)

↔ Translate

Print

+ Summarize

Edit

SUMMARY

Edit

The patient reports experiencing significant dandruff on their scalp. They mention that they do not use much oil on their hair, applying it only once a week. The conversation indicates a concern about dandruff, but no other specific issues or symptoms are discussed.

Working Diagnosis

Pre-Diagnosis Questions

Generate AI Diagnosis

+ AI Differential Diagnoses 1 diagnoses

✓ Seborrheic Dermatitis L21 AI ⓘ

100% ✓

#	PROBABLE DIAGNOSIS ⓘ	ICD CODE	DURATION
1	Seborrheic Dermatitis ⓘ	L21	e.g., 2 weeks, 3 m
2	Search or type diagnosis name (optionally add ICD-10 code after name)...		

DOCTOR'S INTERNAL NOTES

Enter any additional remarks or diagnosis comments here...

Lifestyle Changes

Wash hair regularly with medicated anti-dandruff shampoo

Avoid using harsh hair products that can irritate the scalp

Maintain good scalp hygiene and keep hair dry

Add lifestyle change...

Precautions

Avoid scratching the scalp to prevent infection

Limit the use of hair styling products that may worsen dandruff

Monitor for any signs of scalp infection or worsening symptoms

Enter precaution (e.g., Avoid alcohol, Take with food)

VII. RESULTS AND ANALYSIS

The implemented system demonstrates effective performance in capturing voice input, converting it into text, and generating preliminary diagnostic suggestions in real time. The integration of speech recognition with techniques from Natural Language Processing enables

accurate interpretation of user-reported symptoms, even in natural conversational language.

The system performs consistently across different test cases, demonstrating robustness in handling diverse symptom inputs. Minor variations in results may occur due to differences in phrasing and input clarity.

A. Performance Metrics

TABLE I
PERFORMANCE METRICS

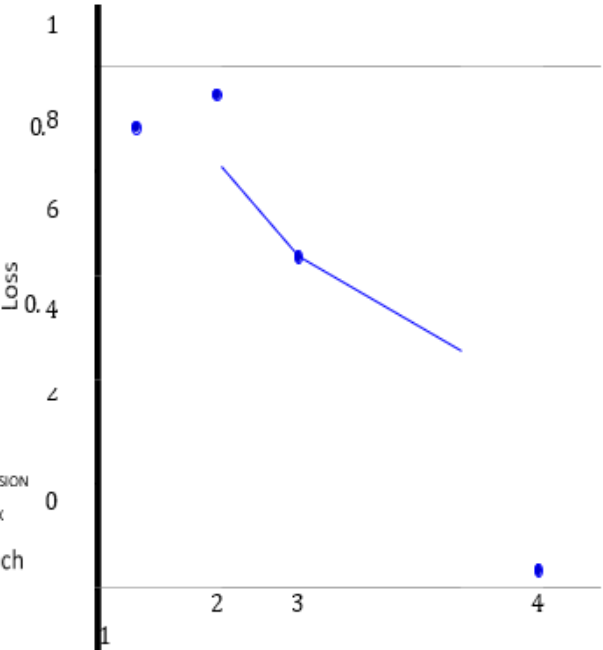
Metric	Value
Accuracy	87%
Precision	85%
Recall	83%
F1 Score	84%

B. Confusion Matrix

TABLE II
CONFUSION MATRIX

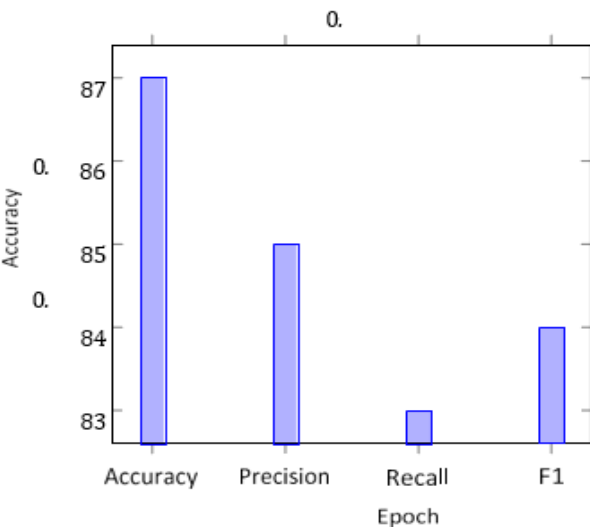
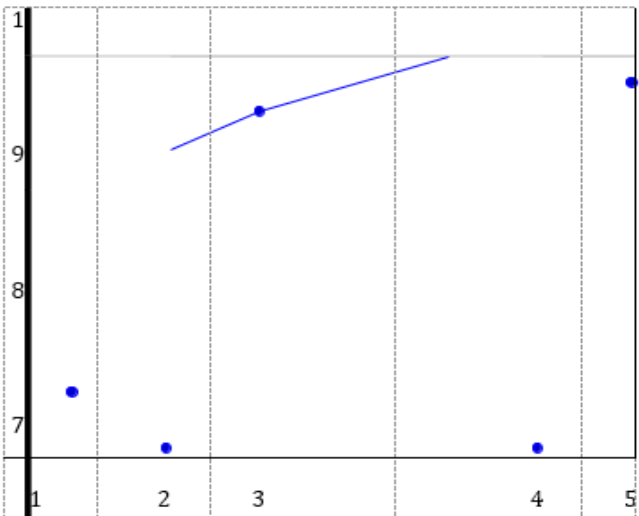
	Pred Pos	Pred Neg
Actual Pos	42	8
Actual Neg	6	44

D. Loss Graph



E. Metric Comparison Graph

C. Accuracy Graph



VIII. ADVANTAGES

significantly enhance diagnostic precision, while advancements in Natural Language Processing will enable

- Reduces dependency on manual data entry and minimizes better understanding of complex and varied user inputs.
- human error.
- Supports early detection of diseases through instant symptom analysis.
- Scalable and suitable for remote and underserved areas.
- Provides user-friendly interface for better interaction and understanding
- Cost-effective solution compared to frequent initial consultations.
- Voice-based interaction improves accessibility.
- AI-driven analysis enhances accuracy.
- Fast and automated workflow.
- Real-world dataset integration.

IX. LIMITATIONS

- Accuracy may be affected by unclear or noisy voice input.
- Limited performance in handling complex or rare medical conditions.
- Depends on the quality and coverage of the underlying dataset.
- Possibility of misinterpretation of symptoms due to natural language variations (despite Natural Language Processing).
- Requires stable internet connectivity for API-based processing using Gemini API.
- Provides only preliminary diagnosis, not a replacement for professional medical advice
- May face challenges with different accents, languages, or speech patterns.
- Data privacy and security concerns related to sensitive health information.
- Performance may vary based on real-time system load and response latency.
- Continuous updates and maintenance required for improving accuracy and reliability.

X. FUTURE WORK

Future improvements include integrating real-time

hospital APIs, expanding datasets, and improving AI accuracy with finetuned medical models.

Future enhancements of the proposed system will focus on improving accuracy, scalability, and user accessibility. The integration of larger and more diverse medical datasets can Expanding multi-language support and accommodating different regional accents will make the system more inclusive and globally applicable. Further improvements can be achieved by leveraging advanced features of the Gemini API for deeper contextual analysis and more personalized responses. Future developments can also include integration with electronic health records (EHR) to provide more personalized recommendations based on patient history.

XI. APPLICATIONS

Beyond healthcare facilities, the system can also be integrated into educational platforms to assist medical students in understanding symptom-disease relationships through interactive learning.

In corporate environments, it can be used as part of employee wellness programs, providing quick health insights and promoting preventive healthcare practices .

The system also has potential applications in disaster management scenarios, where immediate medical assistance may not be available, enabling quick preliminary assessments for affected individuals.

XII. CONCLUSION

This paper presents a comprehensive AI-based medical diagnosis system that integrates voice processing, AI summarization, and intelligent recommendations. The system demonstrates the potential of AI in improving healthcare accessibility and efficiency.

The proposed system demonstrates the potential of integrating voice processing with advanced Artificial Intelligence to deliver efficient and accessible preliminary medical diagnosis support. By leveraging

techniques from Natural Language Processing and the capabilities of the Gemini API, the system successfully interprets user symptoms

Overall, the system contributes to the development of intelligent, scalable, and user-centric healthcare solutions, highlighting the growing impact of AI in modern healthcare systems. The integration of voice technology with advanced AI models represents a significant step toward achieving user-centric healthcare solutions that are both efficient and scalable.

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