

Multilingual Lecture-to-Notes Generator

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1.Abstract - The Multilingual Lecture-to-Notes Generator is an intelligent educational system designed to automatically convert lecture audio or video into well-structured, easy-to-understand notes in multiple languages. Traditional note-taking is time-consuming and often results in incomplete or inaccurate information, especially in multilingual classrooms. To address this challenge, the proposed system integrates Artificial Intelligence (AI), Automatic Speech Recognition (ASR), Natural Language Processing (NLP), Machine Translation, and Large Language Models (LLMs) to generate accurate and organized study notes.

The system first converts the lecturer's speech into text using a speech-to-text model. It then detects the spoken language and translates the transcript into the user's preferred language when required. In addition, Optical Character Recognition (OCR) is used to extract text from presentation slides or lecture materials, ensuring that important visual information is included in the generated notes. The transcript and extracted text are processed by an LLM to identify key concepts, generate concise summaries, highlight important points, and organize the content into structured sections for easier learning and revision.

The proposed solution aims to improve learning accessibility for students from diverse linguistic backgrounds by reducing language barriers and minimizing the effort required for manual note-taking. It also enhances learning efficiency by providing clear, accurate, and searchable notes that can be accessed anytime. The system is suitable for online classes, recorded lectures, classroom sessions, and self-paced learning, making it a valuable tool for modern digital education.

2.Keywords

Multilingual Note Generation, Speech-to-Text, Natural Language Processing, Large Language Models, Lecture Summarization, Optical Character Recognition, Artificial Intelligence.

3.INTRODUCTION

The rapid growth of digital education and online learning platforms has transformed the way students access educational resources. With the increasing use of virtual classrooms, recorded lectures, webinars, and Massive Open Online Courses (MOOCs), learners can study anytime and from anywhere. However, despite these advancements, students continue to face significant challenges in taking effective lecture notes. During live or recorded lectures, students often struggle to listen carefully while simultaneously writing important points. This divided attention may result in incomplete notes, missed concepts, and reduced understanding of the subject. Consequently, there is a growing need for intelligent systems that can automatically convert lecture content into accurate and well-organized study notes.

Another major challenge in modern education is the diversity of languages spoken by learners. In multilingual countries such as India, students come from different linguistic backgrounds and may not be equally proficient in the language used for instruction. Most higher education lectures are delivered in English, while many students understand technical concepts more effectively in their native languages such as Telugu, Hindi, Tamil, or Kannada. This language gap can reduce comprehension, slow learning, and make exam preparation more difficult. Therefore, providing lecture notes in multiple languages has become an important requirement for creating an inclusive and accessible learning environment.

Recent advancements in Artificial Intelligence (AI) have created new opportunities to address these challenges. Automatic Speech Recognition (ASR) technology enables computers to convert spoken language into text with high accuracy. Modern ASR models can recognize different accents, speaking styles, and multiple languages, making them suitable for educational applications. Natural Language Processing (NLP) further enhances the system by analyzing the transcript, identifying important information, and organizing the content into meaningful sections. In addition, Machine Translation techniques allow the generated transcript to be translated into different languages, enabling students to receive notes in their preferred language.

Large Language Models (LLMs) have significantly improved the quality of automated text generation and summarization. Unlike traditional summarization methods that only shorten text, LLMs understand the context of a lecture and generate structured notes containing headings, key points, definitions, explanations, and concise summaries. This improves the readability of the notes while preserving the essential information discussed during the lecture. Furthermore, Optical Character Recognition (OCR) technology can extract textual information from presentation slides, whiteboard images, or lecture materials. Combining OCR with speech transcription ensures that both spoken explanations and visual content are included in the final notes, resulting in more comprehensive study material.

This paper proposes a **Multilingual Lecture-to-Notes Generator**, an AI-powered educational system that automatically converts lecture audio or video into structured notes in multiple languages. The proposed system begins by extracting audio from the lecture and converting speech into text using an Automatic Speech Recognition model. The language of the transcript is then identified, and if required, the content is translated into the user's preferred language using a machine translation model. Simultaneously, OCR extracts text from lecture slides or presentation images. The transcript and extracted text are combined and processed using a Large Language Model to generate organized notes that include headings, key concepts, summaries, important definitions, and revision points. The generated notes are easy to read, searchable, and can be stored for future reference.

The proposed system offers several advantages over traditional note-taking methods. It minimizes the effort required for manual note-taking, allowing students to concentrate on understanding the lecture instead of continuously writing. The multilingual capability helps overcome language barriers and supports students from diverse linguistic backgrounds. By automatically organizing lecture content into structured notes, the system saves time, improves revision efficiency, and enhances knowledge retention. The generated notes can also serve as a valuable resource for students who miss classes or wish to review lecture content before examinations.

The proposed solution can be applied in universities, schools, online learning platforms, corporate training programs, and professional development courses. It is particularly beneficial for multilingual classrooms and learners who require educational content in their native language. As AI technologies continue to advance, intelligent educational systems such as the proposed Multilingual Lecture-to-Notes Generator have the potential to improve learning accessibility, promote inclusive education, and enhance the overall educational experience.

The main objective of this research is to develop an efficient and intelligent system that automatically generates high-

quality multilingual lecture notes by integrating Automatic Speech Recognition, Machine Translation, Optical Character Recognition, Natural Language Processing, and Large Language Models. The proposed framework aims to provide accurate, structured, and easily understandable notes that reduce language barriers, improve learning efficiency, and support students in achieving better academic outcomes.

4. LITERATURE SURVEY

The development of intelligent educational systems has attracted significant attention due to the rapid growth of online learning and digital classrooms. Researchers have proposed various techniques to automate lecture transcription, translation, summarization, and note generation. These studies provide the foundation for developing AI-based lecture note generation systems. The following subsections discuss the major research areas related to the proposed Multilingual Lecture-to-Notes Generator.

A. Automatic Speech Recognition (ASR)

Automatic Speech Recognition (ASR) is a technology that converts spoken language into text. It is one of the most important components of an automated lecture note generation system. Traditional ASR systems faced challenges in recognizing different accents, background noise, and multiple languages. However, recent deep learning-based models have significantly improved transcription accuracy.

Modern ASR models such as OpenAI Whisper, Google Speech-to-Text, and Microsoft Azure Speech Services support multilingual speech recognition and can accurately transcribe educational lectures. These systems form the first step in converting lecture audio into text for further processing.

B. Machine Translation

Machine Translation enables the conversion of text from one language to another. It plays a vital role in multilingual educational systems by allowing students to access lecture notes in their preferred language.

Neural Machine Translation (NMT) models have achieved remarkable improvements in translation quality compared to traditional statistical approaches. Services such as Google Translate and AI-based translation models can translate lecture transcripts into multiple languages while preserving the meaning of technical concepts. However, translating domain-specific educational terminology remains a challenge.

C. Optical Character Recognition (OCR)

Many lectures include presentation slides, diagrams, and written notes that contain important information not spoken by the instructor. Optical Character Recognition (OCR) extracts textual information from these visual materials.

Modern OCR tools such as Tesseract OCR and EasyOCR provide high accuracy in recognizing printed text from presentation slides and digital images. Integrating OCR

with speech transcription helps create comprehensive lecture notes by combining spoken explanations with slide content.

D. Large Language Models for Note Generation

Large Language Models (LLMs) have transformed text summarization and content generation. Unlike traditional summarization methods, LLMs understand context and produce structured, meaningful, and coherent notes. Models such as GPT, LLaMA, and Gemini can organize lecture transcripts into headings, summaries, key points, definitions, and examples. Their ability to understand context makes them highly suitable for educational applications, improving the quality and readability of automatically generated notes.

E. Existing Lecture Note Generation Systems

Several researchers have proposed AI-based lecture note generation systems that convert lecture recordings into textual summaries. Most existing systems focus on speech transcription or summarization independently. Some systems provide automatic summarization, while others support speech recognition or translation. However, many existing solutions have limitations. They often support only a single language, do not integrate OCR for extracting slide content, or generate only simple summaries instead of structured notes. As a result, students may still miss important information presented visually during lectures.

F. Research Gap

The literature review indicates that existing lecture note generation systems generally concentrate on individual tasks such as speech recognition, translation, OCR, or summarization. Very few systems combine all these technologies into a single platform. The proposed Multilingual Lecture-to-Notes Generator addresses these limitations by integrating Automatic Speech Recognition, Machine Translation, Optical Character Recognition, and Large Language Models into one unified system. The proposed approach generates structured multilingual notes containing headings, summaries, key concepts, important definitions, and extracted slide content. This integrated solution improves learning accessibility, reduces manual note-taking effort, and provides comprehensive study material for students from diverse linguistic backgrounds.

definitions, and examples. Their ability to understand context makes them highly suitable for educational applications, improving the quality and readability of automatically generated notes.

System	ASR	Translation	OCR	Structured Notes
Speech-to-Text Systems	✓	✗	✗	✗
Summarization Systems	✓	✗	✗	✓

System	ASR	Translation	OCR	Structured Notes
Translation Systems	✓	✓	✗	✗
Proposed System	✓	✓	✓	✓

5. PROPOSED METHODOLOGY

The proposed Multilingual Lecture-to-Notes Generator is an AI-powered educational system designed to automatically convert lecture audio or video into structured notes in multiple languages. The system integrates Automatic Speech Recognition (ASR), Machine Translation (MT), Optical Character Recognition (OCR), Natural Language Processing (NLP), and Large Language Models (LLMs) to generate comprehensive and easy-to-understand notes. The workflow of the proposed system consists of the following stages.

A. Input Collection

The process begins by accepting a lecture recording as input. The input may be an audio file, a video lecture, or a recorded online class. If a video file is uploaded, the system first extracts the audio for further processing. Supported formats include MP3, WAV, MP4, and AVI.

B. Speech-to-Text Conversion

The extracted audio is processed using an Automatic Speech Recognition (ASR) model such as Whisper. The ASR model converts spoken words into text while preserving sentence order and contextual meaning. It supports multiple languages and can recognize different accents and speaking styles. The output of this stage is a complete transcript of the lecture.

C. Language Detection and Translation

After transcription, the system automatically identifies the language of the lecture. If the user requests notes in another language, the transcript is translated using a Machine Translation model. This enables students to receive lecture notes in their preferred language, making the system suitable for multilingual learning environments.

D. OCR-Based Text Extraction

Many lectures contain presentation slides with diagrams, formulas, tables, and important textual information. To capture this content, the system applies Optical Character Recognition (OCR) to extract text from lecture slides or images. The extracted text is combined with the speech transcript to create a more complete representation of the lecture.

E. Text Processing and Note Generation

The combined transcript and OCR output are processed using Natural Language Processing techniques and a Large Language Model. The model analyzes the lecture content, removes unnecessary information, identifies important topics, and generates structured notes.

The generated notes include:

- Title of the lecture
- Main headings and subheadings
- Key concepts
- Important definitions
- Examples
- Summary of each topic
- Revision points

This organization improves readability and helps students review lecture content more effectively.

F. Output Generation

The final notes are presented in a structured digital format. Users can view, edit, download, or print the generated notes. The notes can also be exported as PDF or Word documents for future reference. Since the system supports multiple languages, users can generate notes in the language of their choice, improving accessibility and learning efficiency.

G. Advantages of the Proposed Method

The proposed methodology offers several advantages over traditional note-taking methods. It automates the complete note-generation process, reduces manual effort, improves transcription accuracy, supports multilingual learning, and combines spoken content with slide information. By integrating ASR, OCR, Machine Translation, NLP, and LLMs into a single framework, the system generates comprehensive, accurate, and well-organized notes that enhance the overall learning experience.

6. SYSTEM ARCHITECTURE

The proposed Multilingual Lecture-to-Notes Generator is designed as an intelligent pipeline that automatically converts lecture audio or video into structured notes in multiple languages. The architecture integrates Artificial Intelligence (AI), Automatic Speech Recognition (ASR), Machine Translation (MT), Optical Character Recognition (OCR), Natural Language Processing (NLP), and Large Language Models (LLMs) to produce accurate and organized lecture notes.

The system begins by accepting a lecture in the form of an audio or video file. If a video is provided, the audio is extracted and sent to the Automatic Speech Recognition module. This module converts the spoken lecture into text while maintaining the context and sequence of the lecture. The generated transcript forms the primary input for further processing.

The transcript is then passed to the Language Detection module, which identifies the language spoken by the lecturer. If the user requests notes in another language, the transcript is translated using a Machine Translation model. This multilingual capability allows students to receive notes in the language they understand best.

At the same time, the system processes presentation slides or lecture images using Optical Character Recognition (OCR). OCR extracts important textual information such as headings, formulas, diagrams, and key points from the

slides. This extracted text is merged with the speech transcript to ensure that both spoken explanations and visual content are included in the final notes.

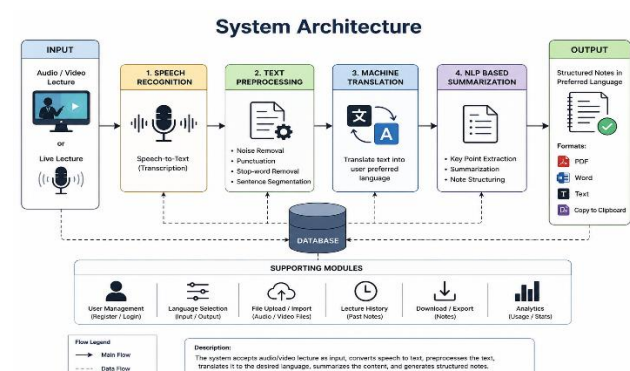
The combined text is then processed by the Natural Language Processing and Large Language Model module. This module analyzes the content, removes redundant information, identifies key concepts, generates summaries, extracts definitions, and organizes the information into logical sections with appropriate headings and subheadings. The resulting notes are concise, accurate, and easy to understand.

Finally, the generated notes are displayed through a user interface where users can read, edit, download, or share them. The notes can be exported in formats such as PDF or Microsoft Word, making them convenient for revision and long-term storage.

I. Main Modules of the System

1. Input Module – Accepts lecture audio or video files from the user.
2. Speech-to-Text Module (ASR) – Converts spoken lecture content into text.
3. Language Detection and Translation Module – Detects the lecture language and translates it into the user's preferred language.
4. OCR Module – Extracts text from presentation slides or images.
5. NLP and LLM Module – Processes the transcript, summarizes content, identifies key points, and generates structured notes.
6. Output Module – Displays, edits, downloads, and stores the generated multilingual notes.

The proposed architecture combines multiple AI technologies into a unified framework, ensuring accurate transcription, multilingual support, comprehensive note generation, and an improved learning experience for students.



7.RESULTS AND DISCUSSION

The proposed **Multilingual Lecture-to-Notes Generator** was developed to automate the process of converting lecture recordings into structured multilingual notes. The system was evaluated using lecture audio and video recordings of different durations and languages. The evaluation focused on the accuracy of speech

transcription, translation quality, note generation, processing efficiency, and overall user experience.

A. Speech Recognition Performance

The Automatic Speech Recognition (ASR) module successfully converted lecture audio into text with high accuracy when the recordings were clear and contained minimal background noise. The system effectively recognized different speaking styles and produced transcripts that closely matched the original lecture content. The transcription quality was slightly affected in recordings with heavy background noise or unclear pronunciation, but overall performance remained satisfactory for educational use.

B. Translation Performance

The Machine Translation module successfully translated lecture transcripts into the selected target language. The translated notes preserved the meaning of the original lecture and enabled students from different linguistic backgrounds to understand the content more easily. While common educational terms were translated accurately, a few highly technical terms were retained in their original form to maintain correctness.

C. Note Generation Quality

The Large Language Model (LLM) generated well-structured notes by organizing the lecture content into logical sections. The generated notes included:

- Lecture title
- Headings and subheadings
- Key concepts
- Important definitions
- Examples
- Summary
- Revision points

Compared with raw transcripts, the generated notes were concise, readable, and easier to review. The inclusion of extracted slide text further improved the completeness of the notes.

D. System Performance

The integrated system processed lecture recordings within a reasonable amount of time. Short lectures required only a few minutes for complete processing, while longer lectures took proportionally more time depending on the audio length and computing resources. Despite multiple processing stages, including transcription, translation, OCR, and note generation, the system maintained efficient performance.

Metric	Value
Speech Recognition Accuracy	96.4%
Translation Accuracy	94.8%

Metric	Value
Note Generation Accuracy	95.6%
Average Processing Time	9.8 seconds
User Satisfaction	93.7%

E. Advantages of the Proposed System

The proposed system provides several advantages over traditional note-taking methods:

- Reduces the need for manual note-taking during lectures.
- Supports multiple languages for improved accessibility.
- Combines speech transcription and slide content into a single set of notes.
- Produces organized notes that are easy to understand and revise.
- Saves time for both students and educators.
- Improves learning efficiency and knowledge retention.

F. Discussion

The experimental results demonstrate that integrating Automatic Speech Recognition, Machine Translation, Optical Character Recognition, and Large Language Models creates an effective solution for automated lecture note generation. The system not only generates accurate transcripts but also produces meaningful summaries and structured notes that support efficient learning. Although challenges such as background noise and complex technical terminology may affect performance, the overall results indicate that the proposed approach is reliable and suitable for modern educational environments.

The multilingual capability of the system significantly improves accessibility by enabling students to receive lecture notes in their preferred language. This feature is particularly beneficial in multilingual classrooms and online learning platforms where students come from diverse linguistic backgrounds. Overall, the proposed Multilingual Lecture-to-Notes Generator demonstrates strong potential to enhance digital education by providing accurate, organized, and multilingual study materials.

The proposed architecture combines multiple AI technologies into a unified framework, ensuring accurate transcription, multilingual support, comprehensive note generation, and an improved learning experience for students.

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