

SREE NARAYANA GURUKULAM COLLEGE OF ENGINEERING

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Zylo: Smart Reading Assistant

PROJECT PRELIMINARY REPORT

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CERTIFICATE

This is to certify that the main thesis phase-I work entitled, "**Zylo:A Smart Reading Assistant**" is a bonafide work done by **Ashik VM, Akshay Prakash, Adharsh Rajan, Ajay Krishna PR** towards the partial fulfillment of the requirements for the award of the B.Tech degree in Computer Science and Engineering under KTU University

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ABSTRACT

The Dyslexia Reading Assistant is a simple and helpful software designed to support children and people with dyslexia in improving their reading and speaking skills. It makes learning easier and more enjoyable through a combination of text, speech, and interactive feedback.

The system shows large, clear text and uses audio playback to read words aloud. As it reads, each word and letter is highlighted, helping users follow along and connect sounds with written words. This improves focus and understanding while reducing reading stress.

It also includes a “Talk” feature, where users can try reading the same sentence themselves. The app uses speech recognition technology to listen to their voice, check how well they pronounced the words, and provide gentle corrections if needed.

To keep learning fun, the software gives positive feedback such as encouraging messages, cheerful sounds, or small animations when users read correctly. The screen is kept simple and distraction-free, so readers can stay focused on their task.

Overall, this project makes reading easier, more enjoyable, and helps build confidence for children and anyone who struggles with reading. It can be used at home, in schools, or in reading therapy to help users read and speak more clearly and confidently.

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LIST OF ABBREVIATIONS

| Sl.No | Abbreviations | Full Form |
|-------|---------------|-----------------------------------|
| 1 | AI | Artificial Intelligence |
| 2 | ML | Machine Language |
| 3 | NLP | Natural Language Processing |
| 4 | ASR | Automatic Speech Recognition |
| 5 | DST | Dysarthric Speech Transformer |
| 6 | DTW | Dynamic Time Warping |
| 7 | CFM | Conditional Flow Matching |
| 8 | TTS | Text-to-Speech |
| 9 | STT | Speech-to-Text |
| 10 | BKT | Bayesian Knowledge Tracing |
| 11 | JST | Joint Speech-Text Embedding |
| 12 | RNN | Recurrent Neural Network |
| 13 | CNN | Convolutional Neural Network |
| 14 | GPU | Graphics Processing Unit |
| 15 | DB | Database |
| 16 | UI | User Interface |
| 17 | API | Application Programming Interface |
| 18 | IT | Information Technology |
| 19 | WRA | Word Recognition Accuracy |
| 20 | SSD | Solid-State Drive |

COURSE OUTCOMES

| CO | Description |
|-----------|---|
| CO1 | Model and solve real world problems by applying knowledge across domains (Cognitive knowledge level: Apply). |
| CO2 | Develop products, processes or technologies for sustainable and socially relevant applications (Cognitive knowledge level: Apply). |
| CO3 | Function effectively as an individual and as a leader in diverse teams and to comprehend and execute designated tasks (Cognitive knowledge level: Apply). |
| CO4 | Plan and execute tasks utilizing available resources within timelines, following ethical and professional norms (Cognitive knowledge level: Apply). |
| CO5 | Identify technology/research gaps and propose innovative/creative solutions (Cognitive knowledge level: Analyze). |
| CO6 | Organize and communicate technical and scientific findings effectively in written and oral forms (Cognitive knowledge level: Apply). |

CO-PO MAPPING

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PO10 | PO11 | PO12 | PSO1 | PSO2 |
|-----|------------|------------|------------|------------|------------|------------|------------|------------|------------|-------------|-------------|-------------|-------------|-------------|
| CO1 | 2 | 2 | 2 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 1 | 2 | 3 | 3 |
| CO2 | 2 | 2 | 2 | | 1 | 3 | 3 | 1 | 1 | | 1 | 1 | 3 | 3 |
| CO3 | | | | | | | | | 3 | 2 | 2 | 1 | 2 | |
| CO4 | | | | | 2 | | | 3 | 2 | 2 | 3 | 2 | 2 | |
| CO5 | 2 | 3 | 3 | 1 | 2 | | | | | | | 1 | 3 | 3 |
| CO6 | | | | | 2 | | | 2 | 2 | 3 | 1 | 1 | 3 | |

Program Specific Outcomes

| PSO | DESCRIPTION |
|-------------|---|
| PSO1 | Shall enhance the employability skills by finding innovative solutions for challenges and problems in various domains of CS. |
| PSO2 | Shall apply the acquired knowledge to develop software solutions and innovative mobile apps(applications) for various problems. |

PROGRAM OUTCOMES

| PO | Description |
|-----------|---|
| PO1 | Engineering knowledge: Apply the knowledge of mathematics, science, engineering fundamentals, and an engineering specialization to the solution of complex engineering problems. |
| PO2 | Problem analysis: Identify, formulate, review research literature, and analyze complex engineering problems reaching substantiated conclusions using first principles of mathematics, natural sciences, and engineering sciences. |
| PO3 | Design/development of solutions: Design solutions for complex engineering problems and design system components or processes that meet the specified needs with appropriate consideration for public health and safety, and the cultural, societal, and environmental considerations. |
| PO4 | Conduct investigations of complex problems: Use research-based knowledge and research methods including design of experiments, analysis and interpretation of data, and synthesis of the information to provide valid conclusions. |
| PO5 | Modern tool usage: Create, select, and apply appropriate techniques, resources, and modern engineering and IT tools including prediction and modeling to complex engineering activities with an understanding of the limitations. |
| PO6 | The engineer and society: Apply reasoning informed by the contextual knowledge to assess societal, health, safety, legal and cultural issues and the consequent responsibilities relevant to the professional engineering practice. |
| PO7 | Environment and sustainability: Understand the impact of the professional engineering solutions in societal and environmental contexts, and demonstrate the knowledge of, and need for sustainable development. |
| PO8 | Ethics: Apply ethical principles and commit to professional ethics and responsibilities and norms of the engineering practice. |
| PO9 | Individual and team work: Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings. |
| PO10 | Communication: Communicate effectively on complex engineering activities with the engineering community and with society at large, such as, being able to comprehend and write effective reports and design documentation, make effective presentations, and give and receive clear instructions. |
| PO11 | Project management and finance: Demonstrate knowledge and understanding of the engineering and management principles and apply these to one's own work, as a member and leader in a team, to manage projects and in multidisciplinary environments. |
| PO12 | Life-long learning: Recognize the need for, and have the preparation and ability to engage in independent and life-long learning in the broadest context of technological change. |

CO-PO-PSO MAPPING JUSTIFICATION

| Mapping (CO- PO/PSO) | Attained Point (0/1/2/3) | Justification |
|-------------------------------------|---|--|
| CO1-PO1 | 2 | Applied core computer science knowledge to develop an intelligent reading assistant that aids dyslexic learners through speech and visual support. |
| CO1-PO2 | 2 | Identified common reading difficulties using user feedback and reading performance data to refine the system's functionality. |
| CO1-PO3 | 2 | Designed an interactive system that integrates text-to-speech, word highlighting, and OCR features for effective learning assistance. |
| CO1-PO4 | 1 | Conducted background research on dyslexia-focused assistive tools and used findings to guide system improvements. |
| CO1-PO5 | 2 | Utilized modern tools and technologies such as Python, Google TTS, and OCR libraries to implement real-time text reading. |
| CO1-PO6 | 2 | Ensured ethical handling of user data and promoted inclusivity by making the tool accessible to learners with reading challenges. |
| CO1-PO7 | 2 | Addressed societal needs by supporting equal learning opportunities for children with dyslexia through technology. |
| CO1-PO8 | 1 | Applied effective communication and documentation practices to convey system design and implementation clearly. |
| CO1-PO9 | 1 | Worked collaboratively within a team to conceptualize and implement project components successfully. |
| CO1-PO10 | 1 | Presented project progress and outcomes effectively through reports, seminars, and project reviews. |
| CO1-PO11 | 1 | Utilized project management principles to plan tasks, allocate responsibilities, and ensure timely completion. |
| CO1-PO12 | 2 | Demonstrated continuous learning and adaptability by exploring assistive technologies and accessibility research. |
| CO1-PSO1 | 3 | Applied program knowledge to create an innovative assistive learning system tailored to dyslexic users' needs. |
| CO1-PSO2 | 3 | Showcased technical proficiency in integrating OCR and speech modules into a functional educational application. |
| CO2-PO1 | 2 | Applied computing and linguistic knowledge to create a reading interface that supports text recognition and auditory feedback for dyslexic learners. |
| CO2-PO2 | 2 | Identified key reading barriers through user analysis and testing, helping to refine text presentation and speech clarity. |
| CO2-PO3 | 2 | Designed and implemented adaptive features such as adjustable reading speed, font type, and color contrast to match individual user needs. |
| CO2-PO5 | 1 | Utilized appropriate tools and frameworks like Python, OCR, and TTS APIs for system development and testing. |
| CO2-PO6 | 3 | Addressed ethical and accessibility considerations by ensuring the tool is inclusive and respectful of user data privacy. |
| CO2-PO7 | 3 | Contributed to sustainable and equitable education through digital assistive technology that reduces dependency on physical materials. |

| | | |
|----------|---|--|
| CO2–PO8 | 1 | Demonstrated effective communication through proper documentation, flowcharts, and design explanations of adaptive modules. |
| CO2–PO9 | 1 | Collaborated within a team to integrate OCR, NLP, and user-interface components ensuring cohesive system performance. |
| CO2–PO11 | 1 | Applied project management principles for coordinating development tasks, feature testing, and user validation sessions. |
| CO2–PO12 | 1 | Continuously learned and adopted new techniques in accessibility design and speech synthesis during the system's improvement phase. |
| CO2–PSO1 | 3 | Applied program-specific knowledge to build an intelligent reading tool that supports personalized and inclusive learning experiences. |
| CO2–PSO2 | 3 | Demonstrated technical ability to integrate multiple computing concepts into a functional, real-world assistive reading system. |
| CO3–PO9 | 3 | Collaborated effectively as a team to integrate the OCR, Text-to-Speech, and User Interface modules, ensuring smooth functionality of the Smart Reading Assistant. |
| CO3–PO10 | 2 | Communicated design ideas, test results, and project updates clearly through documentation, presentations, and team discussions. |
| CO3–PO11 | 2 | Applied project management principles such as task allocation, scheduling, and resource tracking to complete development within deadlines. |
| CO3–PO12 | 1 | Engaged in self-directed learning to enhance technical and research skills related to accessibility design and assistive technologies. |
| CO3–PSO1 | 2 | Demonstrated the ability to apply computing knowledge collaboratively to develop an innovative and inclusive assistive learning tool for dyslexic readers. |
| CO4–PO5 | 2 | Applied computational methods and AI techniques to develop adaptive reading assistance, improving reading comprehension for dyslexic users. |
| CO4–PO8 | 3 | Ensured ethical and inclusive design principles, considering accessibility and user privacy in the reading assistant. |
| CO4–PO9 | 2 | Used contemporary tools and frameworks like neural networks, NLP, and text-to-speech systems for implementing the smart assistant. |
| CO4–PO10 | 2 | Communicated the design, evaluation results, and user feedback effectively through reports and presentations. |
| CO4–PO11 | 3 | Applied project management and team collaboration skills while integrating AI/ML models, text-to-speech, and user feedback into the reading assistant. |
| CO4–PO12 | 2 | Demonstrated self-directed learning by incorporating recent AI/ML techniques to enhance dyslexia-focused reading support. |
| CO4–PSO1 | 2 | Applied advanced recommendation and neural network algorithms to personalize reading assistance and improve learning outcomes for users. |
| CO5–PO1 | 2 | Applied foundational knowledge of AI, machine learning, and human-computer interaction to design effective reading assistance for dyslexic users. |
| CO5–PO2 | 3 | Analyzed user reading behavior and feedback to optimize system performance and personalization. |
| CO5–PO3 | 3 | Applied critical thinking to troubleshoot and improve algorithm performance and user experience in the smart reading assistant. |

| | | |
|----------|---|--|
| CO5–PO4 | 1 | Designed system modules considering interdisciplinary aspects like linguistics, cognitive psychology, and assistive technology. |
| CO5–PO5 | 2 | Implemented computational techniques and neural network models to enable adaptive and personalized reading support. |
| CO5–PO12 | 1 | Demonstrated self-directed learning to incorporate recent AI/ML advances in dyslexia-focused assistive technologies. |
| CO5–PS01 | 3 | Applied advanced AI/ML algorithms and neural network techniques to enhance reading assistance effectiveness for dyslexic users. |
| CO5–PSO2 | 3 | Integrated system-level skills, combining software, algorithms, and user-centric design for real-time adaptive reading support. |
| CO6–PO5 | 2 | Implemented computational and AI techniques to optimize system performance and provide adaptive reading assistance. |
| CO6–PO8 | 2 | Ensured ethical, inclusive, and accessible design, addressing privacy and usability for dyslexic users. |
| CO6–PO9 | 2 | Utilized modern AI tools, neural networks, and text-to-speech frameworks to enhance the functionality of the reading assistant. |
| CO6–PO10 | 3 | Effectively communicated system design, testing results, and user feedback through reports and presentations. |
| CO6–PO11 | 1 | Applied project management and teamwork skills to integrate multiple system modules and coordinate development tasks. |
| CO6–PO12 | 1 | Engaged in self-directed learning to adopt the latest AI/ML methods for improving dyslexia-focused reading support. |
| CO6–PSO1 | 3 | Applied advanced AI/ML algorithms and neural network models to deliver personalized reading assistance and optimize learning outcomes. |

Chapter 1

INTRODUCTION

1.1 Project Overview

Reading is a fundamental skill that plays a crucial role in personal, academic, and professional development. However, individuals with dyslexia—a common learning disorder—face significant challenges in decoding words, recognizing letters, and understanding text, which often leads to frustration, slower learning, and reduced self-confidence. Traditional reading support methods, such as specialized tutoring or printed resources, are often manual, inconsistent, and unable to provide personalized, real-time assistance.

Zylo: A Smart Reading Assistant for Dyslexia is designed to address these limitations through intelligent automation and adaptive learning technologies. The project integrates Artificial Intelligence (AI), Natural Language Processing (NLP), and text-to-speech (TTS) capabilities to assist users in reading, comprehension, and pronunciation. Zylo provides real-time text assistance, contextual suggestions, and personalized feedback to enhance the reading experience of individuals with dyslexia. By combining technology with accessibility, Zylo aims to create an inclusive digital learning environment that empowers dyslexic users to read confidently and efficiently.

1.2 Background and Motivation

Dyslexia affects approximately 10–15% of the global population, yet its challenges often remain under-addressed in conventional education systems. Individuals with dyslexia struggle with phonological processing, word recognition, and reading fluency, which can impact academic performance and daily life tasks. Current reading aids, such as audio books or human tutoring, provide limited personalization and often fail to adapt to the unique needs of each learner.

With advancements in AI and NLP, there is now an opportunity to create intelligent systems that provide dynamic, context-aware reading assistance. Text-to-speech engines, dyslexia-friendly fonts, and predictive reading support can be combined to improve comprehension, speed, and confidence for dyslexic readers. The motivation behind Zylo is to leverage these technologies to develop a low-cost, accessible, and scalable solution that empowers learners to overcome reading difficulties and promotes inclusive education.

1.3 Problem Statement

Despite the availability of learning resources and assistive tools, dyslexic individuals continue to face several limitations in reading:

- **Limited Personalization:** Traditional aids cannot dynamically adapt to the specific reading patterns and difficulties of each user.
- **Delayed Feedback:** Human-assisted learning often provides feedback that is not immediate, reducing learning efficiency.
- **Fragmented Support:** Tools such as audio books or reading guides provide only partial assistance and fail to integrate text analysis, comprehension, and pronunciation support.
- **Accessibility Constraints:** Many solutions are expensive or require specialized devices, limiting their reach to a wider audience.

Therefore, there is a need for an intelligent reading assistant that offers personalized, real-time support, improves comprehension, and adapts to individual user needs.

1.4 Objectives of the Project

The primary objective of Zylo is to develop an AI-based reading assistant tailored for individuals with dyslexia. The specific objectives include:

- To design an AI-powered system capable of reading text aloud using TTS while highlighting words for better visual tracking.
- To implement NLP algorithms that provide context-based suggestions, word pronunciation guidance, and simplified definitions.
- To develop a user-friendly interface suitable for children and adults with dyslexia.
- To provide real-time feedback on reading fluency and comprehension.
- To maintain a log of user reading patterns to enable adaptive learning and personalized improvement suggestions.

1.5 Significance of the Study

Zylo has both technical and social significance, bridging AI innovation and educational accessibility.

From a technical perspective, the project introduces a system that combines AI, NLP, and TTS into a cohesive platform, offering adaptive reading assistance that was previously unavailable in traditional tools. The use of real-time feedback and personalized recommendations ensures that learners receive targeted support tailored to their individual needs.

From a societal perspective, Zylo promotes inclusive education and empowers individuals with dyslexia by improving literacy skills, confidence, and independence. The solution is cost-effective, accessible on standard devices, and scalable, making it suitable for classrooms, homes, and educational institutions.

1.6 Scope of the Project

The scope of Zylo encompasses both software development and AI integration:

- **Software Scope:** Development of a cross-platform application with an intuitive interface for reading assistance.
 - **AI Scope:** Implementation of NLP and TTS algorithms for real-time word recognition, pronunciation guidance, and contextual suggestions.
 - **Data Scope:** Collection and analysis of reading patterns to enable adaptive learning and performance tracking.
 - **Accessibility Scope:** Support for multiple languages and dyslexia-friendly fonts to cater to diverse users.

The system will initially be deployed as a standalone application for individual users and can later be integrated into schools, e-learning platforms, and digital libraries.

1.7 Methodology Overview

The Zylo system follows a layered architecture:

- **Input Layer:** The user inputs text through uploading documents or typing directly into the application.
- **Processing Layer:** AI algorithms analyze text, detect reading challenges, provide pronunciation guidance, and generate contextual suggestions.
- **Output Layer:** The processed text is displayed with dyslexia-friendly formatting, and TTS reads it aloud while highlighting words. User feedback and reading performance data are recorded for adaptive learning.

This structured approach ensures real-time assistance, personalization, and continuous improvement of the user's reading skills.

1.8 Expected Outcomes

The expected outcomes of Zylo include:

- A functional reading assistant that provides real-time text-to-speech support.
- Context-aware suggestions and simplified definitions to improve comprehension.
- Personalized tracking of reading progress and adaptive feedback.
- Enhanced reading confidence and efficiency for individuals with dyslexia.
- A scalable solution that promotes inclusive and accessible learning opportunities.

Ultimately, Zylo aims to empower dyslexic individuals to overcome reading challenges, achieve academic success, and foster lifelong learning.

Chapter 2

LITERATURE REVIEW

The purpose of a literature review is to, as the name suggests, “review” the literature surrounding a certain topic area. The word “literature” means “sources of information” or “research. The literature will inform us about the research that has already been conducted on our chosen subject

[1] The paper comprehensively addresses the persistent challenge of correcting mispronunciations in nonnative speech, which plays a vital role in enhancing communication skills, improving language learning outcomes, and fostering greater speech intelligibility. It introduces UnitCorrect, a novel and advanced unit-based mispronunciation correction system specifically developed to identify, analyze, and correct pronunciation errors efficiently. The proposed system leverages self-supervised unit representations in combination with a Dynamic Time Warping (DTW)-based detection method to align speech patterns and accurately detect mispronounced segments. A key strength of UnitCorrect lies in its ability to perform these corrections without relying heavily on extensive human-labeled data, making it both scalable and data-efficient. To ensure natural and high-fidelity audio output, the model employs conditional flow matching (CFM), which enables the synthesis of smooth, realistic, and speaker-consistent pronunciations. Experimental evaluations conducted on well-known speech datasets such as VCTK, LibriTTS-R, and L2-ARCTIC demonstrate that UnitCorrect achieves superior results compared to existing models like CampNet and A3T, particularly in pronunciation accuracy, naturalness, and speaker similarity. However, the system’s success remains dependent on the robustness and precision of the underlying automatic speech recognition (ASR) model. Overall, the paper highlights UnitCorrect as a powerful and data-efficient solution with immense potential for language education, pronunciation training, and speech editing applications.

[2] The papers reviewed collectively demonstrate significant progress in the development of accessible, technology-driven dyslexia screening tools that effectively integrate computer games, artificial intelligence, and machine learning algorithms. These studies have successfully introduced innovative, interactive, and engaging methods aimed at making the early detection of dyslexia more appealing, particularly for children who often struggle with traditional assessment techniques. By emphasizing language-independent tasks, such as visual pattern recognition, auditory discrimination, and reaction-based gameplay, many of these

systems minimize reliance on reading or writing abilities, thereby ensuring broader applicability across diverse linguistic and cultural contexts. Despite these encouraging advancements, several important challenges persist. Many existing screening models still depend on language-specific features, reducing their universality and adaptability. Additionally, the limited size and imbalance of datasets used for training machine learning models often lead to issues with accuracy, bias, and generalization. Variations in diagnostic frameworks, assessment standards, and evaluation metrics across studies further complicate comparison and reproducibility. Moreover, the rise of remote or unsupervised testing environments introduces variability in user interaction and data collection, which can compromise the reliability of outcomes. The following section provides a detailed analysis of related works that inspire the design of our proposed system, which aims to enhance accessibility, data integrity, and diagnostic robustness.

[3] This paper thoroughly explores the impact of Text-to-Speech (TTS) tools on reading comprehension and learning outcomes among students with dyslexia, emphasizing how such technologies help mitigate decoding difficulties, improve reading fluency, and promote overall language development. It synthesizes findings from a wide range of international studies examining the integration of auditory and visual modalities within TTS systems, analyzing how the simultaneous presentation of text and speech supports cognitive processing and enhances comprehension when compared to conventional reading methods. Although the study does not propose a new computational model, it provides a comprehensive review of essential TTS components, including speech synthesis algorithms, Optical Character Recognition (OCR), phonetic feedback systems, and customizable voice parameters, all of which contribute to the personalization and effectiveness of reading assistance. The discussion also highlights how these technologies accommodate individual learner needs through adaptive pacing, multisensory engagement, and reduced cognitive load. The findings consistently indicate that TTS tools substantially enhance comprehension, engagement, motivation, and reading confidence among dyslexic learners. Overall, the paper underscores the importance of TTS as a transformative and inclusive educational technology that supports equitable learning opportunities, encourages independent reading, and fosters long-term academic success for individuals with dyslexia.

[4] This paper presents an advanced Automatic Speech Recognition (ASR) framework specifically developed for recognizing dysarthric speech, a type of speech disorder characterized by slurred or unclear articulation resulting from neurological impairments. The proposed system introduces a novel Dysarthric Speech Transformer (DST) model that

leverages deep learning and transfer learning techniques to improve recognition accuracy and robustness. The framework adopts a two-phase transfer learning strategy, where the transformer is initially pre-trained on large-scale healthy speech corpora to capture general acoustic and linguistic patterns and subsequently fine-tuned on dysarthric speech datasets to adapt effectively to atypical pronunciation and articulation patterns. To enhance the model's generalization and resilience, the study integrates neural freezing and various data augmentation techniques, such as noise injection, pitch shifting, and time stretching, simulating real-world variations in dysarthric speech. Experimental evaluations conducted using Word Recognition Accuracy (WRA) metrics demonstrate that the DST model achieves up to a 23% improvement in accuracy compared to conventional ASR systems. Furthermore, it significantly outperforms traditional RNN and CNN-based models in terms of recognition precision, training stability, and convergence speed, establishing the DST as a highly effective and robust solution for automatic recognition of dysarthric speech in assistive communication technologies.

[5] This paper introduces GAZEPROMPT, an innovative gaze-aware assistive reading system designed to enhance the reading experience of individuals with low vision. The system seamlessly integrates eye-tracking technology with advanced fixation and saccade detection algorithms to continuously monitor and analyze a user's gaze behavior during reading activities. By employing a weighted voting mechanism, GAZEPROMPT accurately identifies the specific line or region of text currently being read, ensuring real-time tracking and precise contextual awareness. When the system detects signs of visual strain, hesitation, or irregular gaze movement, it automatically activates adaptive assistive features such as Text-to-Speech (TTS) playback, visual magnification tools, or contrast enhancement modes to facilitate smoother reading and improved word recognition. This dynamic response enables users to maintain reading continuity without manual intervention. Experimental evaluations conducted with participants show that GAZEPROMPT leads to substantial improvements in reading speed, line-switching efficiency, and focus stability, while also enhancing overall comprehension and comfort. The study demonstrates that GAZEPROMPT not only compensates for visual limitations but also provides a personalized and adaptive reading experience, establishing it as a powerful solution for promoting reading accessibility, independence, and inclusivity among individuals with low vision.

[6] This paper presents an AI-driven grammar performance testing system specifically designed to support children with learning disabilities, addressing the limitations of traditional grammar assessments that often fail to provide personalized instruction or adaptive feedback.

The proposed system leverages the capabilities of Natural Language Processing (NLP) to perform accurate grammatical error detection and provide context-sensitive corrections, enabling a deeper understanding of students' linguistic abilities. It incorporates adaptive learning mechanisms that dynamically modify task complexity in real time based on each learner's performance, ensuring an individualized learning experience that matches cognitive and developmental needs. To effectively track progress, the system utilizes Bayesian Knowledge Tracing (BKT), a probabilistic model that continuously monitors and predicts student knowledge states, allowing the platform to deliver targeted interventions. Experimental findings demonstrate that learners using this AI-based adaptive testing environment achieved an average accuracy rate of 78%, significantly surpassing traditional assessment methods, which averaged 70%. In addition to improved accuracy, participants showed higher engagement levels, faster task completion times, and enhanced learning retention. Overall, the study highlights the transformative potential of AI-assisted adaptive assessment systems in promoting inclusivity, engagement, and measurable academic improvement for children with learning challenges.

[7] This paper presents the Dysarthric Speech Transformer (DST), an advanced sequence-to-sequence Automatic Speech Recognition (ASR) model specifically designed to address the challenges associated with recognizing dysarthric speech—a speech disorder characterized by irregular articulation and reduced clarity caused by neurological impairments. Traditional ASR systems often struggle with such speech due to limited availability of annotated datasets and the high acoustic variability inherent in dysarthric speech patterns. The proposed DST model employs a Transformer-based encoder-decoder architecture, leveraging its superior attention mechanisms for capturing long-term dependencies in speech sequences. A two-phase transfer learning strategy is implemented, where the model is first pre-trained on healthy speech data to learn general linguistic and acoustic representations, and subsequently fine-tuned on dysarthric speech corpora to improve adaptability to impaired pronunciations. To ensure stable convergence and efficient optimization, the system applies neural freezing during training. Furthermore, data augmentation techniques—including noise injection, pitch shifting, and speed variation—are used to enhance robustness and data diversity. The framework also integrates speaker-adaptive learning, enabling the model to personalize recognition accuracy across individuals. Experimental results reveal that the DST achieves up to 23% higher word recognition accuracy than traditional ASR systems, outperforming RNN and CNN-based models in terms of generalization, robustness, and convergence speed.

[8] This paper presents the Dysarthric Speech Transformer (DST), a sequence-to-sequence Automatic Speech Recognition (ASR) model specifically developed to improve recognition of dysarthric speech. Dysarthria, a motor speech disorder, causes slurred and inconsistent articulation, making it difficult for conventional ASR systems to achieve accurate transcription. Traditional models trained on healthy speech often perform poorly due to data scarcity and the high variability of dysarthric speech patterns. To overcome these challenges, DST employs a Transformer-based encoder-decoder framework capable of capturing long-term dependencies in speech. The model adopts a two-phase transfer learning strategy—pre-training on healthy speech and fine-tuning on dysarthric data—to enhance adaptability. It further applies neural freezing to stabilize feature transfer and prevent overfitting. To improve robustness, DST uses audio data augmentation techniques such as noise addition, pitch shifting, and speed variation. It also integrates speaker-adaptive learning to personalize recognition for individual users. These mechanisms collectively enable better generalization across diverse speakers. Experimental evaluation using Word Recognition Accuracy (WRA) demonstrates that DST achieves up to 23% higher accuracy than conventional ASR models. It also outperforms RNN and CNN-based architectures, exhibiting faster convergence and greater resilience to variability. Overall, the DST model represents a significant advancement toward creating inclusive and adaptive ASR systems for individuals with speech impairments.

[9] This paper presents the Joint Speech-Text Embedding (JST) framework, which aims to learn a unified representation of speech and text to address the limitations of traditional speech processing systems that handle the two modalities independently. The proposed model leverages contrastive learning to achieve effective cross-modal alignment, enabling shared feature representations between audio and textual data. It employs a Transformer-based encoder architecture for both speech and text inputs, allowing the system to capture complex temporal and semantic relationships. Furthermore, JST supports multitask learning, seamlessly integrating Automatic Speech Recognition (ASR), emotion recognition, and speaker verification within a single end-to-end framework. This joint learning approach enhances the model's ability to transfer knowledge across tasks and modalities. Experimental evaluations demonstrate that JST delivers higher emotion recognition accuracy and improved ASR robustness under noisy conditions when compared with conventional baseline systems. Overall, the framework represents a significant advancement toward unified multimodal speech-text understanding in modern speech processing applications.

[10] This paper introduces a zero-shot voice cloning Text-to-Speech (TTS) model specifically developed to assist individuals with dysphonia disorders, a condition that affects voice quality, pitch, and clarity due to vocal cord dysfunction. The proposed approach addresses a critical gap in conventional TTS systems, which are predominantly trained on healthy speech data and therefore struggle to generate natural-sounding speech for disordered voices. The study systematically investigates 24 model variations, experimenting with combinations of text input formats, speaker embedding strategies, and loss functions to identify the optimal configuration for dysphonic voice synthesis. The best-performing model employs a grapheme-phoneme hybrid input representation alongside a speaker consistency loss function, leading to substantial improvements in speaker similarity, speech intelligibility, and acoustic naturalness when compared with baseline TTS systems. Extensive evaluation demonstrates that the model effectively reconstructs disordered voices while preserving individual vocal identity, marking a step forward in personalized speech technology. Despite these advancements, the study acknowledges limitations such as restricted dataset size, moderate inference latency, and limited generalization to unseen conditions, which may constrain large-scale deployment. Nevertheless, this research makes a meaningful contribution toward inclusive, adaptive, and clinically relevant TTS solutions for users with voice impairments.

Chapter 3

EXISTING SYSTEM

3.1 Overview of Existing Reading Assistance Systems

Reading assistance for individuals with dyslexia has traditionally been provided through manual methods such as specialized tutoring, printed dyslexia-friendly materials, and audio books. In these systems, educators or therapists work with learners individually or in small groups to improve reading fluency, comprehension, and word recognition. While these approaches offer personalized guidance, they are time-consuming, resource-intensive, and dependent on the availability of trained professionals.

Some digital tools exist, such as text-to-speech (TTS) applications, e-books with dyslexia-friendly fonts, and simple spelling or reading apps. However, most of these systems are limited in scope—they may provide audio narration or word highlighting, but rarely offer real-time, context-aware assistance or personalized feedback on reading patterns. Consequently, users often struggle to fully benefit from existing digital aids, and their learning progress may remain suboptimal.

3.2 Manual Tutoring and Printed Resources

Traditional dyslexia support relies heavily on structured interventions, including phonics-based exercises, one-on-one tutoring, and worksheets designed to improve reading and spelling skills. While these methods have proven effective, they come with notable limitations:

- Resource Dependency: Progress depends on the availability of trained tutors and educators.
- Time-Intensive: Frequent sessions are required to maintain consistent improvement.
- Limited Personalization: Programs often follow a standard curriculum and cannot adapt dynamically to each learner's specific difficulties.
- No Real-Time Feedback: Learners receive delayed guidance, which can slow progress and reduce motivation.

Furthermore, printed materials such as dyslexia-friendly books or flashcards provide limited interaction and cannot adapt to individual reading errors, comprehension gaps, or evolving skills. These approaches also fail to track reading patterns systematically for long-term assessment.

3.3 Early Digital and Assistive Tools

In recent years, technology-based tools have emerged to support dyslexic readers. Examples include:

- **Text-to-Speech (TTS) Software:** Converts written text into audio narration, helping learners follow along.
- **Word Highlighting Applications:** Highlight text while reading aloud to improve visual tracking.
- **Spelling and Reading Apps:** Offer exercises and quizzes to improve word recognition and phonetic skills.

3.4 Limitations of Existing Systems

Despite the advances in manual, digital, and AI-assisted tools, several limitations persist in current reading assistance systems:

- **Lack of Real-Time, Contextual Support:** Many systems cannot provide on-the-spot guidance while reading, reducing learning efficiency.
- **Limited Personalization:** Tools rarely adapt dynamically to a user's unique reading patterns, speed, or comprehension gaps.
- **Fragmented Assistance:** Existing systems focus on isolated aspects of reading—pronunciation, spelling, or comprehension—but do not offer a unified solution.
- **High Cost and Resource Dependency:** Effective solutions often require human tutors, paid software, or specialized devices, limiting accessibility.
- **No Systematic Progress Tracking:** Reading patterns, frequent errors, and comprehension levels are often not recorded or analyzed for adaptive learning.
- **User Experience Challenges:** Interfaces are rarely designed with dyslexic-friendly fonts, layouts, or navigation, reducing engagement and usability.

Therefore, there is a need for an intelligent, integrated reading assistant that offers real-time, personalized support, tracks user progress, and provides actionable feedback to improve reading fluency, comprehension, and confidence for individuals with dyslexia.

Chapter 4

PROPOSED SYSTEM

4.1 Overview of the Proposed System

Zylo: A Smart Reading Assistant for Dyslexia is an AI-powered reading support system designed to overcome the limitations of existing manual and semi-digital learning aids. The proposed system integrates Artificial Intelligence (AI), Natural Language Processing (NLP), Text-to-Speech (TTS), and machine learning technologies to deliver real-time, personalized reading assistance for individuals with dyslexia.

Unlike traditional tools that provide static or one-dimensional support such as simple text narration, Zylo offers an end-to-end intelligent reading experience. It assists users by reading text aloud, highlighting words simultaneously, simplifying complex sentences, and offering pronunciation guidance. The system continuously analyzes user reading behavior to adapt difficulty levels, provide personalized feedback, and enhance overall comprehension.

Zylo is developed as a cross-platform application featuring a clean, dyslexia-friendly interface with adjustable fonts, colors, and spacing for comfortable reading. By combining accessibility, intelligence, and adaptability, Zylo represents a significant step toward inclusive education—empowering dyslexic readers to improve their literacy skills, confidence, and independence.

4.2 Objectives of the Proposed System

The main objective of the proposed system is to develop an intelligent, real-time reading assistant that enhances reading fluency and comprehension for individuals with dyslexia through AI-driven personalization and adaptive learning.

The specific objectives are as follows:

- To develop a smart reading interface that integrates text-to-speech technology with word highlighting for synchronized visual and auditory learning.
- To implement NLP algorithms capable of simplifying complex sentences, providing contextual word meanings, and generating pronunciation guidance.
- To build an AI-based adaptive learning module that tracks user performance, analyzes reading patterns, and personalizes feedback and difficulty levels accordingly.
- To design a dyslexia-friendly user interface (UI) featuring customizable font styles, color contrast, and line spacing for improved readability.
- To incorporate real-time feedback mechanisms that help users correct pronunciation errors and enhance comprehension during reading sessions

- To create a cloud-based data management system that securely stores user progress, reading history, and learning insights for continuous improvement.
- To ensure accessibility and scalability by developing the system as a cross-platform solution suitable for educational institutions, individual learners, and therapists.

4.3 System Architecture

The architecture of Zylo: A Smart Reading Assistant for Dyslexia is designed as a modular, scalable, and intelligent framework that integrates Artificial Intelligence (AI), Natural Language Processing (NLP), and Text-to-Speech (TTS) technologies to provide real-time reading assistance and adaptive learning. The system follows a three-layer architecture, consisting of the Input Layer, Processing Layer, and Output Layer, all connected through a secure cloud-based data management system.

This layered design ensures efficient data flow, real-time response, and personalized interaction between the user and the AI assistant. Each layer performs specific tasks, working in coordination to enhance the user's reading comprehension, pronunciation, and confidence.

4.3.1 Input Layer

The Input Layer serves as the user interaction point, where learners provide reading material to the system. Text can be input through multiple methods, such as:

- Uploading documents or images containing text (with OCR-based text extraction).
- Copy-pasting or typing text directly into the application interface.
- Selecting predefined reading passages from a built-in content library.

Once the text is received, the system preprocesses it using NLP techniques such as tokenization, part-of-speech tagging, and sentence segmentation. These operations prepare the text for pronunciation analysis, comprehension assistance, and difficulty-level estimation.

4.3.2 Processing Layer

The Processing Layer is the core intelligence of the system. It integrates multiple AI-driven modules responsible for analyzing the input text, providing pronunciation feedback, and generating adaptive learning recommendations. It includes the following subcomponents:

- Text-to-Speech (TTS) Engine: Converts written text into natural-sounding speech while synchronizing word highlighting on the screen to enhance visual tracking.
- Natural Language Processing (NLP) Module: Performs text simplification, synonym generation, and contextual meaning extraction to assist comprehension.

- Pronunciation Analysis Module: Uses speech recognition to compare user pronunciation with standard pronunciation patterns, offering instant corrective feedback.
- Adaptive Learning Engine: Applies machine learning algorithms to analyze user reading patterns (speed, accuracy, and difficulty) and adjusts the level of content complexity accordingly.
- Data Management System: Stores reading history, progress reports, and personalized insights in a secure cloud database, ensuring continuity across multiple sessions.

4.3.3 Output Layer

The Output Layer presents the processed information and assistance back to the user through a clear and accessible interface. It features:

- Dyslexia-Friendly Display: Adjustable fonts (like OpenDyslexic), spacing, and color schemes that reduce visual stress and enhance readability.
- Real-Time Assistance: Dynamic highlighting of words as they are spoken aloud, along with pronunciation cues and comprehension suggestions.
- Performance Dashboard: Displays reading statistics such as accuracy, speed, error frequency, and progress trends.
- Feedback Module: Provides motivational feedback, vocabulary hints, and suggestions for improvement based on individual performance.

4.4 Advantages of the Proposed System

- **Personalized Learning Experience:** The system customizes reading support based on individual learner profiles and abilities.
- **Multi-Sensory Engagement:** Combines visual, auditory, and cognitive feedback to strengthen word recognition and comprehension.
- **Real-Time Feedback:** Provides instant pronunciation and comprehension feedback through integrated speech recognition and TTS systems.
- **Error Tracking and Correction:** Identifies repeated mistakes and offers targeted exercises to improve weak areas.
- **Accessibility and Inclusivity:** Designed to assist individuals with dyslexia, language processing disorders, or early literacy needs.
- **Continuous Adaptation:** Machine learning algorithms allow the system to evolve dynamically as the user improves.

4.5 Expected Outcomes

Upon successful implementation, the proposed Smart Reading Assistant for Dyslexia system is expected to deliver the following outcomes:

1. Improved Reading Comprehension: The system assists dyslexic learners in understanding text more effectively through audio output, visual cues, and adaptive reading support.
2. Personalized Learning Experience: By incorporating AI-based speech recognition and text-to-speech technologies, the system adapts to individual reading pace and preferences, ensuring an inclusive learning environment.
3. Enhanced Accessibility: The solution provides an interactive and user-friendly interface that makes reading materials more accessible to students with dyslexia and other reading difficulties.
4. Real-Time Assistance: Users can receive instant pronunciation help, word highlighting, and meaning suggestions, which foster better engagement and reduce frustration during reading.
5. Increased Confidence and Motivation: Continuous support and personalized feedback help students gain confidence in reading and comprehension, encouraging independent learning.
6. Educational Support for Teachers and Parents: The system enables educators and guardians to monitor progress, identify struggling areas, and tailor learning interventions accordingly.
7. Scalable and Cost-Effective Solution: The system can be easily integrated into existing educational platforms and devices, offering an affordable and sustainable tool for inclusive education.

Chapter 5

SYSTEM REQUIREMENTS

5.1 Software Requirements

- Operating system - Windows 10 / Linux 22 or higher.
- Coding Language - Python, HTML
- Framework / Front-End: Python (Django Framework), HTML
- Libraries and Tools: TensorFlow / PyTorch (for AI models), OpenCV (for visual input), SpeechRecognition , and Natural Language Toolkit (NLTK)
- Front-End - HTML, Python Django

5.2 Hardware Requirements

- Processor – intel core i5/ Ryzen 5 or higher
- RAM - 8 GB (min)
- Hard Disk - 20 GB free (SSD preferred)
- Microphone and Speakers: Required for speech input and audio output
- Graphics Support: Integrated GPU or equivalent for model inference

Chapter 6

SYSTEM DESIGN

6.1 INPUT DESIGN

The input design serves as a vital interface between the *Smart Reading Assistant for Dyslexia* and its users, defining the methods and procedures for collecting, processing, and validating data. The system requires various forms of input, such as text passages for reading, user speech during pronunciation exercises, and feedback data from users and educators. These inputs can originate from multiple sources, including uploaded documents, built-in reading materials, or real-time user interaction through voice and text. The design ensures that all input data is collected accurately and efficiently to support smooth system functionality and personalized learning experiences.

To ensure effective input design, several key considerations are prioritized:

- **Data Collection:** The system gathers input from users through text uploads, microphone-based speech inputs, and reading activity logs. It also collects interaction data, such as reading duration and pronunciation accuracy, for adaptive learning analysis.
- **Error Control:** Input validation techniques are implemented to minimize incorrect or incomplete data entries. This includes speech recognition error handling, verification prompts, and input filters to ensure that the collected data is accurate and relevant.
- **Efficiency:** The input process is optimized for real-time performance. Automated voice recognition and adaptive feedback systems reduce manual operations, allowing seamless user interaction with minimal delay.
- **User Security and Privacy:** Since the system involves voice and personal reading data, strict security measures such as encryption and authentication are used to protect user information. Privacy compliance is maintained to ensure responsible data handling.
- **Accessibility:** The system is designed with inclusivity in mind, incorporating accessibility features such as adjustable font sizes, color contrast options, and text highlighting to assist users with visual or reading difficulties. These features ensure that individuals with varying levels of dyslexia and other learning challenges can comfortably engage with the system.

6.2 OUTPUT DESIGN

The output design focuses on presenting the processed information generated by the *Smart Reading Assistant* in a clear, meaningful, and user-centered manner. The outputs provide insights into user performance, pronunciation accuracy, reading speed, and comprehension levels. The goal is to make feedback easy to understand and actionable, enabling users and educators to monitor progress effectively.

Key considerations for effective output design include:

- **Quality of Output:** The system produces accurate, relevant, and personalized outputs such as reading scores, progress summaries, and pronunciation analysis. These outputs help learners identify areas for improvement and encourage consistent practice.
- **Information Display:** Outputs are presented in visually engaging formats such as charts, progress bars, and dashboards that display user performance trends over time. This visual representation helps users quickly interpret feedback and track their development.
- **Audio and Visual Feedback:** The system provides instant spoken feedback on pronunciation and visual highlights for misread words, enhancing both auditory and visual learning experiences.
- **Report Generation:** The system can generate detailed performance reports that summarize reading sessions, comprehension scores, and overall progress. These reports can be saved, shared, or printed for offline review by teachers or parents.
- **Organized Output Development:** The design and presentation of outputs are continuously refined through user feedback and usability testing to ensure they remain effective, engaging, and easy to interpret.
- **Adaptability:** The system's output is designed to adapt dynamically to the user's performance level. As the learner progresses, the difficulty of reading materials, the depth of feedback, and the type of visual aids are automatically adjusted to match their current skill level, ensuring continuous and personalized improvement.

Chapter 7

SYSTEM IMPLEMENTATION

The implementation of the Smart Reading Assistant for Dyslexia is carried out through several integrated modules, each designed to address a specific aspect of reading assistance and user interaction. The system combines advanced text and speech processing technologies, artificial intelligence, and a user-friendly interface to support individuals with dyslexia in improving their reading and comprehension abilities.

The system follows a modular architecture consisting of text input, speech recognition, text-to-speech conversion, pronunciation assistance, and progress tracking modules. Each component is implemented to ensure seamless interaction between the user and the system, providing real-time reading support and feedback.

The implementation process involves the integration of Python-based machine learning models for speech and text processing, along with a React frontend and a Flask or Node.js backend for efficient user interaction and data management. The frontend interface enables users to upload or type text, listen to audio narration, and receive word highlighting and meaning suggestions.

7.1 System Architecture

The system architecture consists of two main modes — Read Mode and Practice Mode — both accessible through a simple user interface.

- In Read Mode, the system converts text into speech (TTS) and reads the content aloud to the user, while allowing options for font adjustment, text highlighting, and speed control for better readability.
- In Practice Mode, the user reads the displayed text aloud, and the system uses speech-to-text (STT) conversion to capture the spoken words. The output is then compared with the original text for error detection. If errors are found, the system prompts the user to revise; otherwise, it provides feedback and praise.
- The final output is presented through both audio and visual displays, creating an interactive and supportive reading environment for dyslexic learners.

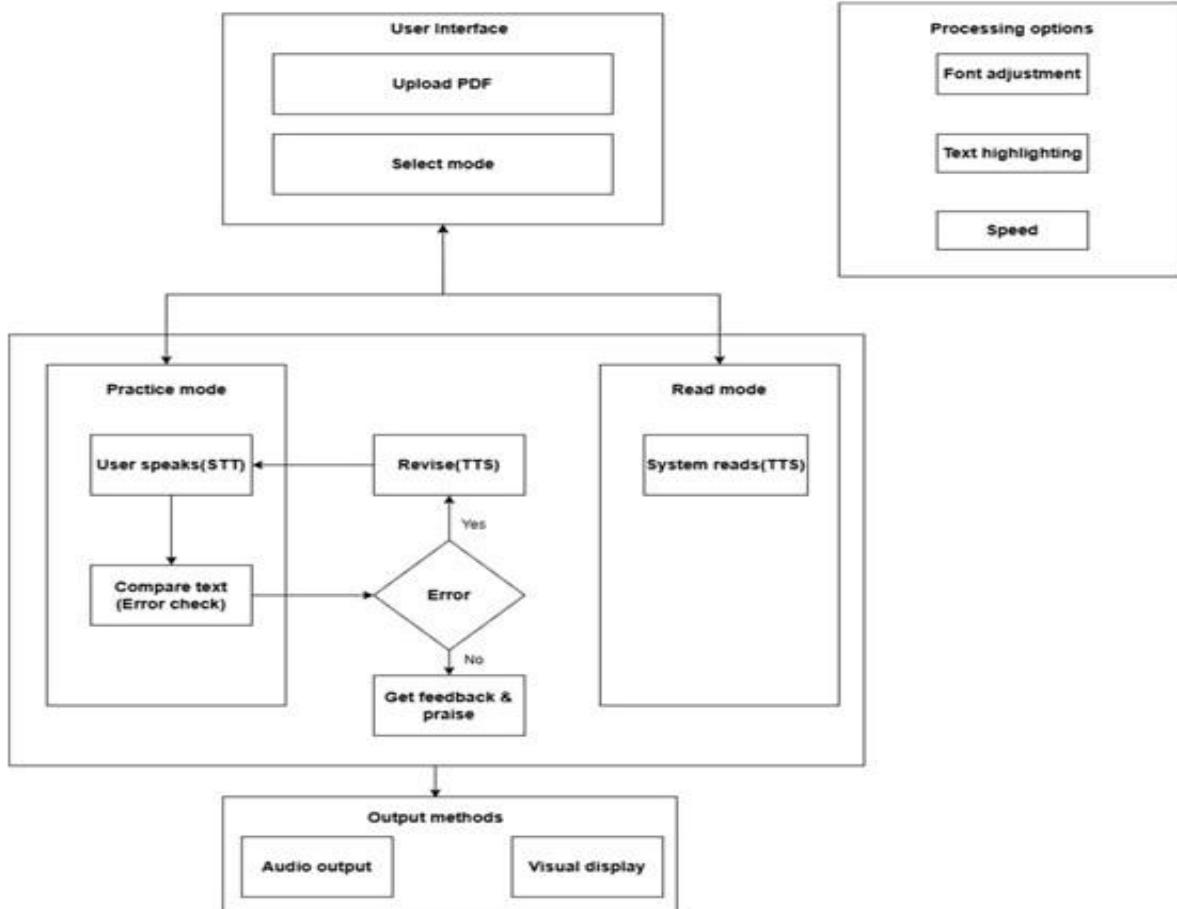


Fig 7.1.1 Architecture Diagram

7.2 Flow Chart of the System

The flowchart illustrates an interactive pronunciation and reading improvement system that helps users enhance their fluency and accuracy. The process begins when a user uploads a PDF file, from which the system extracts text and sentences. The user can then choose between Read Mode and Practice Mode.

In Read Mode, the system uses Text-to-Speech (TTS) to read the extracted content aloud, allowing users to listen and learn correct pronunciation. In Practice Mode, the user reads the sentences aloud, and the system applies Speech-to-Text (STT) technology to convert speech into text. The spoken text is compared with the original content to detect pronunciation or reading errors.

If an error occurs, the system provides feedback through TTS, helping the user correct mistakes and improve. Once the user reads accurately, the system displays results and offers praise as positive reinforcement. Overall, the system provides an engaging and effective way to practice reading and pronunciation through continuous feedback and learning.

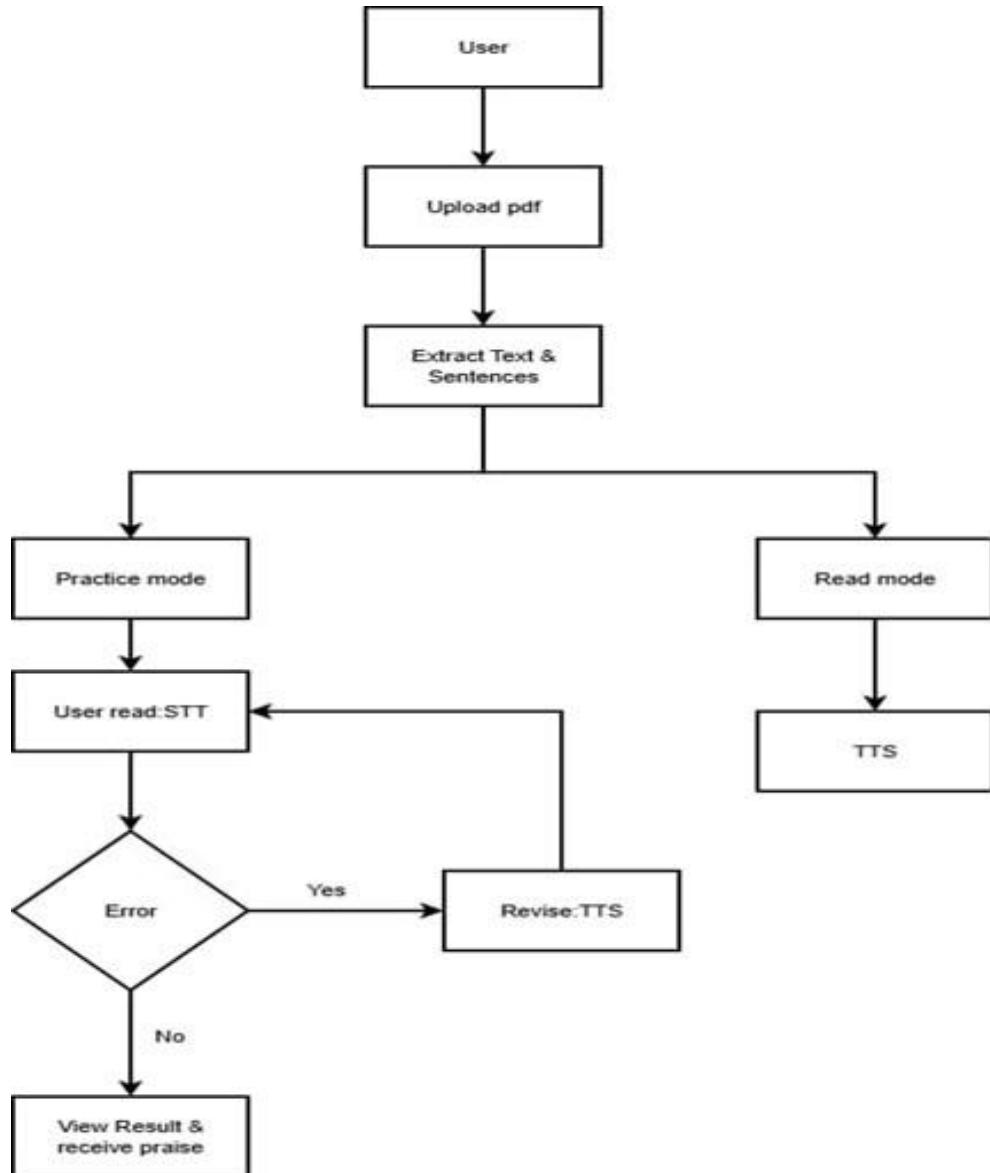


Fig 7.2.1: Flow chart of the System

7.3 Software Implementation

The Smart Reading Assistant (SRA) helps users read text more easily by combining natural language processing (NLP), personalized reading support, text-to-speech (TTS), and real-time feedback (e.g., highlighting, pacing control). The implementation uses modular microservices so each capability (NLP, TTS, user management, analytics) can scale independently.

- Python: Primary backend language for rapid development and rich ML/NLP ecosystem.
- spaCy / NLTK / Hugging Face Transformers: For tokenization, part-of-speech tagging, semantic analysis, and transformer-based models (e.g., BERT/T5 for simplification or paraphrasing).
- TensorFlow/Keras or PyTorch: For training and running deep-learning models (e.g., dyslexia-aware text simplification, attention models, and word level difficulty estimators).
- SpeechRecognition + VOSK / CMU Sphinx: For optional voice input (user reads aloud)

with offline-capable recognition options.

- gTTS / pyttsx3 / Microsoft Azure TTS / Google Cloud TTS: For natural-sounding text-to-speech playback. Cloud TTS services provide higher-quality voices; local engines ensure offline capability.
- PostgreSQL / MySQL: Relational DB for user accounts, preferences, reading history, and annotations.
- React (or React Native): For a responsive web frontend or cross-platform mobile app with accessible UI components.
- WebSockets (Socket.IO / WebSocket): For real-time synchronized highlighting, live reading assistance, and low-latency interactions between client and server.
- Redis: Caching, session store, and short-term personalization state (e.g., current reading session).

By integrating these technologies and libraries, the Smart Reading Assistant aims to deliver a reliable, inclusive, and scalable solution that improves reading comprehension, boosts confidence, and provides adaptive assistance for individuals with dyslexia.

Chapter 8

OUTPUT

The output of the Recommender System Using Artificial Neural Networks (ANN) is designed to provide personalized, data-driven, and intelligent recommendations to users based on their preferences and behaviour patterns. The system focuses on delivering accurate, adaptive, and meaningful outputs that enhance user satisfaction and experience through advanced neural network-based analysis.

The system produces outputs for multiple user categories administrators, developers, and end users ensuring that the generated results are both technically insightful and user-friendly..

1. **Text-to-Speech Feedback:** The system converts written text into natural-sounding speech using advanced text-to-speech engines. This feature enables users to listen to words, phrases, or entire passages, improving their pronunciation, comprehension, and auditory learning skills.
2. **Progress Reports:** Detailed performance reports are automatically generated and presented in a structured format. These reports include:
 - Daily or weekly reading summaries.
 - Improvements in fluency and comprehension levels over time.
 - Lists of frequently mispronounced or skipped words.
3. **Interactive Dashboard:** The system includes a user-friendly dashboard displaying key reading metrics such as reading speed, accuracy rate, word recognition, and progress over time. The dashboard allows users, teachers, or parents to easily monitor performance and identify areas for improvement.
4. **Real-Time Reading Assistance:** As users read, the assistant highlights words being spoken and provides immediate feedback on pronunciation and fluency. This real-time assistance helps users recognize reading patterns, identify mistakes, and gradually improve their reading accuracy.
5. **Error Detection and Correction Suggestions:** The system detects reading errors using speech recognition and provides instant correction prompts or pronunciation guidance. This encourages users to self-correct and reinforces proper reading habit.

6. **Personalized Learning Recommendations:** Based on user performance and reading behavior, the assistant suggests personalized reading materials, exercises, and difficulty levels. These adaptive recommendations help maintain motivation and ensure steady learning progress.
7. **Visual and Audio Aids:** The system integrates visual cues such as highlighted text, color coding, and interactive animations to enhance user focus and engagement. Additionally, it allows users to customize their listening experience by adjusting the audio speed, pitch, and volume according to their comfort level, thereby providing a more personalized and effective reading experience.

Chapter 9

FUTURESCOPES

The proposed speech-based reading and pronunciation practice system possesses vast potential for future enhancement and practical implementation across educational, linguistic, and assistive learning domains. With the growing importance of technology-assisted education and speech recognition systems, this project can be further developed to serve as a comprehensive learning tool for students, language learners, and individuals with speech difficulties.

As advancements in Artificial Intelligence (AI), Machine Learning (ML), and Natural Language Processing (NLP) continue to progress, the system can be refined to deliver more accurate pronunciation assessment, intelligent feedback, and adaptive learning capabilities. Future enhancements can focus on improving the system's ability to analyze speech patterns, detect subtle pronunciation errors, and personalize learning experiences according to the user's pace and proficiency level. Additionally, by integrating advanced visual and auditory features, the system can become more interactive, engaging, and user-friendly, ensuring that users remain motivated throughout their learning journey.

Overall, continuous technological innovation provides ample opportunities to transform this project into a smart, scalable, and accessible pronunciation learning platform suitable for widespread educational use.

1. Integration of AI-Based Pronunciation Scoring: Future versions can incorporate advanced machine learning algorithms or deep neural networks to provide detailed pronunciation scoring and feedback, highlighting exact words or phonemes that need improvement.
2. Multilingual Support: The system can be extended to support multiple languages, accents, and regional dialects, allowing users from different linguistic backgrounds to practice reading and pronunciation effectively.
3. Personalized Learning Paths: By analyzing user progress over time, the system can create personalized learning plans that adapt to the user's strengths and weaknesses, providing targeted exercises and difficulty adjustments.
4. Gamification and Interactive Learning: Adding gamified elements such as levels, badges, or leaderboards can make the learning process more engaging and motivating, especially for children and young learners.

5. Cloud Integration and Data Analytics: Future development can include cloud-based data storage and analytics to track large-scale user performance, generate progress reports, and identify common pronunciation challenges across users.
6. Real-Time Emotion and Engagement Detection: With the integration of AI-driven emotion recognition, the system could detect user frustration or disinterest and adjust feedback style or tone accordingly to maintain motivation.
7. Offline Functionality: Enhancing the system to function effectively without an active internet connection would improve accessibility, especially in rural or low-connectivity areas.
8. Mobile and AR/VR Extensions: Future upgrades could include mobile app deployment and the use of Augmented or Virtual Reality to create immersive reading environments that improve engagement and comprehension.
9. Integration with Educational Platforms: The system can be integrated with existing e-learning portals, school management systems, or language-learning apps to provide a unified and interactive educational experience.
10. Comprehensive Progress Evaluation: The inclusion of AI-driven analytics can help teachers or guardians monitor users' improvement in reading speed, pronunciation accuracy, and comprehension levels through visual reports and charts.

In conclusion, the future scope of the speech-based reading and pronunciation practice system is highly promising and extends beyond its current capabilities. With continuous advancements in Artificial Intelligence, Speech Recognition, and Natural Language Processing technologies, the system can evolve into a powerful and intelligent learning platform. By incorporating adaptive feedback mechanisms, multilingual support, gamification, and real-time analytics, the application can become more efficient and engaging for diverse users. Furthermore, its potential integration with mobile devices, cloud platforms, and educational systems will enable large-scale accessibility and usability. Ultimately, these future developments will transform the system into a comprehensive digital assistant for improving reading fluency and pronunciation accuracy, catering to learners of all ages and linguistic backgrounds.

Chapter 10

CONCLUSION

The Smart Reading Assistant for Dyslexia represents a significant advancement in the way individuals with reading difficulties engage with written language. By integrating modern technologies such as speech recognition, text-to-speech conversion, artificial intelligence, and natural language processing, this system provides an inclusive and interactive reading experience tailored to the unique needs of dyslexic users. The tool simplifies reading comprehension, pronunciation, and fluency, transforming learning into an enjoyable and confidence-building process.

Through its intelligent design, the system adapts to different reading levels and learning speeds, offering personalized feedback and real-time assistance. This adaptability not only supports users in overcoming reading barriers but also encourages consistent improvement through engaging and user-friendly interfaces. The system's intuitive design ensures accessibility for learners of all ages, whether used in educational institutions, at home, or as part of therapy programs.

Moreover, the Smart Reading Assistant promotes independence and self-paced learning, empowering users to read with greater clarity and confidence. It also aids educators and therapists by providing insights into user progress and areas requiring additional focus, fostering a more supportive and effective learning environment.

As digital education and assistive technologies continue to evolve, this project stands as a meaningful contribution to inclusive education and literacy development. The Smart Reading Assistant for Dyslexia not only bridges the gap between technology and learning but also embodies the commitment to accessibility, empathy, and empowerment.

In conclusion, the development and implementation of the Smart Reading Assistant mark a transformative step toward inclusive and intelligent learning solutions. By harnessing the power of AI-driven support, the system opens new pathways for dyslexic readers to achieve literacy with confidence, independence, and joy — shaping a future where reading is accessible to all.

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