

Mobile Application for Identifying Dyslexia in Children

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(CS)

SESSION 2021-2025

*A thesis submitted to The University of Agriculture, Peshawar in partial fulfillment
of the requirement for the degree of*



BACHELOR OF SCIENCE IN COMPUTER SCIENCE

SUPERVISED BY

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INSTITUTE OF COMPUTER SCIENCES AND INFORMATION TECHNOLOGY

FACULTY OF MANAGEMENT AND COMPUTER SCIENCES

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Table of Contents

Acknowledgment	7
Chapter 1: Introduction	8
1.1 Project Introduction	8
1.2 Problem Statement	9
1.3 Objectives	11
1.4 Background and Motivation	12
Chapter 2: Literature Review	14
2.1 Introduction	14
2.2 Understanding Dyslexia: A Deeper Look	14
• Before School	15
• School Age	15
• Teens and Adults	16
2.3 Traditional Methods of Dyslexia Screening	17
• The Dyslexia Early Screening Test (DEST)	17
• The Wechsler Intelligence Scale for Children (WISC)	17
• The Comprehensive Test of Phonological Processing (CTOPP)	18
2.4 The Rise of Mobile Screening Tools	18
• Dyslexia Quest	19
• Dytective	19
• Nessy Learning	19
• Ghotit Real Writer	19
2.5 Research Gaps and Opportunities	20
• Identified Gaps	20
• Our Opportunity	21
2.6 Summary	21
Chapter 3: Methodology	22
3.1 System Overview	22
3.2 UI/UX Prototyping	23
3.3 Frontend Development using Flutter	23

• Why Flutter?	23
• Application Overview	23
• Architecture and State Management	24
• Features and Functionalities	24
• Security and Data Privacy	26
• Integration with Backend and AI Models	26
• Development Tools and Environment	26
3.4 AI Model Development	27
3.5 Backend Architecture	27
3.6 Test Result Submission and Retrieval	28
3.7 Admin Dashboard APIs	28
3.8 Integration Between Backend and AI Model	28
System Flow Diagram	29
 Chapter 4: Implementation Phase	30
4.1 UI/UX Prototyping Using Figma	30
4.2 Frontend Implementation (Mobile Application)	33
• User Interface Development	33
• Authentication Module	36
• Test Module Integration	37
• Real-Time Updates	43
4.3 Backend Implementation	43
• Authentication and Role Management	43
• Test Result Handling	44
• AI Model Communication	44
4.4 Database Design	45
4.5 AI Model Implementation	45
• Data Handling	45
• Model Training	45
• Prediction Workflow	46
• Evaluating Results	45
• Deployment	45
• Limitation	46

4.6 Security Measures	46
4.7 Testing and Debugging	47
Chapter 5: Testing and Evaluation	48
5.1 Testing Objectives	48
5.2 Testing Environment	48
5.3 Testing Methodologies	49
• Functional Testing	49
• Usability Testing	50
• Integration Testing	50
5.4 AI Model Accuracy Testing	51
5.5 Security Testing	51
5.6 Summary of Testing Results	51
Chapter 6: Results and Discussion	53
6.1 Overview of Results	53
6.2 Functional Performance Discussion	53
6.3 AI Model Effectiveness	54
6.4 Usability and User Experience	54
6.5 Integration and System Reliability	54
6.6 Limitations Observed	55
6.7 Overall Discussion	55
Chapter 7: Conclusion	56
7.1 Project Summary	56
7.2 Distinctive Features	56
7.3 Impact and Outcomes	56
7.4 Future Scope	57

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CHAPTER 1: INTRODUCTION

1.1 Project Introduction:

Dyslexia is a widely recognized yet frequently misunderstood learning disability that primarily affects an individual's ability to read, write, and spell with fluency and accuracy. It is a neurological condition that impacts the brain's language processing ability, making it difficult for individuals to identify speech sounds and relate them to letters and words.

Dyslexia is not a sign of low intelligence or lack of effort many individuals with dyslexia possess average or above-average intelligence and may excel in areas such as creativity, problem-solving, or visual thinking. However, if not diagnosed and addressed at an early stage, dyslexia can lead to significant academic setbacks, emotional distress, and social difficulties, especially during a child's formative years.

Although dyslexia is one of the most common learning disabilities, affecting an estimated 5–10% of school-aged children globally, it often goes undetected, particularly in low-resource settings. Early detection is critical, as it allows for timely interventions, including specialized teaching techniques, assistive technologies, and academic accommodations that can greatly improve learning outcomes. Without early identification and intervention, children with dyslexia are more likely to experience repeated academic failures, which can result in poor self-esteem, anxiety, and long-term educational disadvantages.

In the context of modern technology, there exists a promising opportunity to address this challenge. The proliferation of smartphones and mobile internet has transformed how information is accessed, and services are delivered across the globe, including in remote and underserved areas. Similarly, advancements in artificial intelligence (AI) and machine learning have enabled computers to analyze large sets of data and identify patterns with remarkable accuracy capabilities that can be harnessed to assist in the early screening of learning disorders such as dyslexia.

This project aims to develop a **mobile application** that leverages **artificial intelligence** to support the early detection of **dyslexia in children**. The application will allow parents, teachers, and caregivers to conduct assessments based on a child's language development, reading ability, spelling performance, and behavioral responses. These inputs will be processed by a trained AI model that can recognize patterns associated with dyslexia and provide a preliminary diagnostic report. The results will be presented in a clear, user-friendly

format and accompanied by recommendations on whether to seek further professional evaluation.

Furthermore, the application will incorporate strong security measures to ensure that all user data is stored and transmitted safely, maintaining the privacy of children and their families. The long-term vision of this project is to democratize access to early screening tools for dyslexia, reducing the barriers posed by cost, distance, and lack of specialized knowledge. By making such tools widely available through a mobile app, the project contributes to a more inclusive and supportive educational environment where every child has the opportunity to thrive.

1.2 Problem Statement:

Dyslexia is a specific learning disorder that significantly affects a child's ability to read, write, spell, and comprehend written language. Despite being common, it remains one of the most underdiagnosed and misunderstood conditions among school-aged children, especially in regions where awareness and resources are limited. Children with dyslexia often face considerable challenges in academic settings, where reading and writing are fundamental skills required for learning across all subjects. When left unidentified, these challenges can escalate, resulting in repeated academic failure, low self-confidence, social withdrawal, and behavioral issues.

Traditional dyslexia diagnosis relies heavily on formal assessments conducted by trained professionals such as educational psychologists, speech-language therapists, or neurodevelopmental specialists. These assessments typically involve standardized tests, clinical interviews, and observations, which although effective, are time-consuming, expensive, and not universally accessible. In many parts of the world, especially in low-income and rural areas, such services are either unavailable or unaffordable, making early detection a luxury rather than a standard.

Moreover, the early symptoms of dyslexia such as difficulty learning letter names, confusing similar-sounding words, or struggling to match sounds with symbols are often mistaken for laziness, carelessness, or general disinterest in learning. This lack of awareness among parents and educators further delays identification and intervention. By the time a child receives a diagnosis, they may have already internalized feelings of failure and frustration, which can deeply affect their academic motivation and mental well-being.

In today's digital age, mobile technology is increasingly accessible and widely used across various socioeconomic groups. Coupled with the advancements in artificial intelligence, mobile platforms offer a powerful means of reaching populations that otherwise lack access to formal educational or psychological support. Yet, despite this technological potential, there remains a scarcity of AI-driven mobile applications specifically designed for dyslexia detection.

This gap presents a critical problem: there is not widely available, user-friendly, and affordable tool that can help non-specialists such as parents and teachers identify early signs of dyslexia in children. Without such tools, countless children continue to struggle without understanding the cause of their difficulties, missing the window of opportunity for early and effective intervention.

Therefore, the central problem this project aims to solve is the lack of accessible, intelligent, and reliable digital solutions for early-stage dyslexia screening. The proposed AI-based mobile application addresses this gap by offering an initial assessment platform that can be used in everyday environments like homes and classrooms, helping to bridge the gap between suspicion and diagnosis and enabling timely support for children in need.

1.3 Objectives:

The primary aim of this project is to design, develop, and evaluate a mobile application that utilizes artificial intelligence to assist in the early detection of dyslexia in children. The application will serve as a preliminary screening tool to help parents, teachers, and caregivers recognize early signs of dyslexia, thereby encouraging timely professional evaluation and intervention.

To achieve this main objective, the following specific objectives have been established:

- 1. To conduct an in-depth study of dyslexia and its early indicators**
Investigate the neurological, cognitive, and behavioral aspects of dyslexia, including symptoms, causes, and developmental patterns. This research will inform the design of the app's assessment criteria and AI model features.
- 2. To identify key assessment indicators and features related to dyslexia**
Develop a framework of measurable parameters such as phonemic awareness, letter-sound recognition, reading fluency, spelling patterns, and short-term memory performance that can be used to evaluate dyslexia-related challenges.
- 3. To design a user-friendly mobile application interface**
Create an intuitive, interactive, and accessible mobile application that allows non-specialist users (such as parents and teachers) to easily input observations and complete assessments without requiring technical or medical expertise.

4. **To develop and train a machine learning model for dyslexia prediction**
Build an AI model using appropriate datasets, including synthetic and real-world samples, that can identify patterns in linguistic and cognitive behavior indicative of dyslexia. The model should achieve acceptable levels of accuracy, sensitivity, and specificity.
5. **To integrate the AI model into the mobile application for real-time prediction**
Ensure seamless integration of the AI backend with the mobile front-end, allowing users to receive immediate feedback based on their input data.
6. **To ensure data privacy and secure storage of sensitive information**
Implement encryption and secure database practices to protect user data, particularly considering that the app will involve sensitive information about children's learning behavior.
7. **To provide clear, comprehensible, and actionable feedback to users**
Present the AI-generated results in simple language with appropriate visual aids and guidance. Offer next steps such as contacting a professional or beginning supportive practices at home or in school.
8. **To evaluate the effectiveness, usability, and impact of the application**
Conduct user testing and performance analysis to assess the application's accuracy, ease of use, and practical value in real-world scenarios. Collect feedback to improve the app's design and reliability

1.4 Background and Motivation:

In every classroom, there are children full of dreams, questions, and imagination. Some easily grasp reading and writing, while others, despite trying just as hard maybe even harder struggle to make sense of letters and words. These children often face confusion, frustration, and sometimes even shame, not knowing why they're "different." One of the most common but still widely misunderstood reasons for this struggle is **dyslexia**.

Dyslexia is a specific learning disorder that primarily affects a person's ability to read and interpret words, letters, and other symbols. It has nothing to do with intelligence. In fact, many people with dyslexia are highly creative and intelligent, yet face difficulties in the traditional school setting due to how their brains process language. This can lead to poor academic performance, low self-esteem, and mental stress, especially when the condition goes unnoticed.

In Pakistan, like in many developing countries, awareness of dyslexia is extremely low. Parents and teachers often mistake it for laziness, disobedience, or a lack of interest in studies. Psychological assessments are rarely available in government schools, and the topic of learning disorders is barely touched upon in teacher training. As a result, countless

children go through school being misunderstood, underperforming, and eventually dropping out without ever knowing they had dyslexia.

As final-year students in Computer Science, we were driven by a simple but powerful question: How can we use technology to help these children, and their parents identify dyslexia early? The answer became the foundation of our project.

We decided to develop a mobile application that can help in the early identification of dyslexia in children, with a special focus on accessibility and ease of use. The goal is not to replace psychologists or clinical diagnosis, but to offer a simple screening tool that can alert parents or teachers when a child might need professional attention. In short, the app acts like a bridge connecting families with the possibility of timely help and support.

Our motivation is deeply personal. Each of us has seen classmates, siblings, or cousins struggle with reading and spelling in ways that didn't make sense at the time. Looking back, some of them may very well have been dyslexic but they were never identified, supported, or understood. This app is our way of contributing to a more inclusive future, where children with dyslexia can be seen, supported, and empowered to reach their full potential.

Chapter 2: Literature Review

2.1 Introduction

To create an effective, relevant, and well-informed solution for identifying dyslexia in children, it is crucial to first understand the research, theories, and technologies that already exist in this field. A literature review is more than just a summary of past studies; it is an opportunity to connect previous knowledge with the goals of our project and to identify the gaps our work aims to fill.

This chapter explores various dimensions of dyslexia: what it is, how it affects learning, and why early detection is so important. It also discusses the limitations of traditional screening methods and the emergence of digital tools particularly mobile applications as a more accessible way to help children who may be struggling with this learning difficulty.

By analyzing both academic research and real-world tools, this chapter builds a strong foundation for our project and justifies the development of a mobile application that can support early identification of dyslexia in an easy, cost-effective, and empowering manner.

2.2 Understanding Dyslexia: A Deeper Look

Dyslexia is one of the most diagnosed learning disorders worldwide, yet it remains deeply misunderstood by the public. According to the International Dyslexia Association (IDA) [1], dyslexia is a specific learning disorder that is neurological in origin. It primarily affects the skills involved in accurate and fluent word reading and spelling, and it is typically characterized by difficulties with phonological processing, which refers to the ability to recognize and manipulate the sound structures of language [2].

This difficulty in processing sounds is what makes reading and writing particularly hard for individuals with dyslexia. For example, a child may confuse letters that look similar, such as "b" and "d", or may struggle to decode new words, even after repeated exposure [3]. While the child may have normal or even above-average intelligence, the effort required to read fluently can be significantly higher than that of their peers.

Symptoms

Signs of dyslexia can be difficult to recognize before your child enters school, but some early clues may indicate a problem. Once your child reaches school age, your child's teacher may be the first to notice a problem. Severity varies, but the condition often becomes apparent as a child starts learning to read [4].

Before school

Signs that a young child may be at risk of dyslexia include:

- Late talking
- Learning new words slowly
- Problems forming words correctly, such as reversing sounds in words or confusing words that sound alike
- Problems remembering or naming letters, numbers and colors
- Difficulty learning nursery rhymes or playing rhyming games

School age

Once your child is in school, dyslexia symptoms may become more apparent, including:

- Reading well below the expected level for age
- Problems processing and understanding what is heard
- Difficulty finding the right word or forming answers to questions
- Problems remembering the sequence of things
- Difficulty seeing (and occasionally hearing) similarities and differences in letters and words
- Inability to sound out the pronunciation of an unfamiliar word
- Difficulty spelling
- Spending an unusually long-time completing tasks that involve reading or writing
- Avoiding activities that involve reading

Teens and adults

Dyslexia signs in teens and adults are a lot like those in children. Some common dyslexia symptoms in teens and adults include:

- Difficulty reading, including reading aloud
- Slow and labor-intensive reading and writing
- Problems spelling

- Avoiding activities that involve reading
- Mispronouncing names or words, or problems retrieving words
- Spending an unusually long-time completing tasks that involve reading or writing
- Difficulty summarizing a story
- Trouble learning a foreign language
- Difficulty doing math word problems

Many researchers have emphasized that early identification is the key to minimizing the long-term impact of dyslexia. A child who is identified and supported at age six has a much higher chance of improving their reading skills than one who is diagnosed at age twelve. According to Shaywitz (2003) [5], timely intervention can “change the trajectory of a child’s life,” while delays can result in academic failure, emotional distress, and social isolation.

Despite being a common condition, dyslexia often goes unnoticed, especially in under-resourced educational systems. The signs are sometimes subtle, and without proper training, teachers may mistake them for behavioral issues or lack of effort [6]. This lack of understanding makes it even more important to build tools that increase awareness and offer simple ways to spot early warning signs [7].

2.3 Traditional Methods of Dyslexia Screening

For many years, identifying dyslexia required a combination of clinical observation, psychological testing, and one-on-one interviews. These methods, while thorough, are also time-consuming, expensive, and dependent on the availability of trained professionals.

Some of the most widely used tools in clinical or academic environments include:

The Dyslexia Early Screening Test (DEST)

The Dyslexia early screening test (DEST) [8] developed by Nicolson and Fawcett, DEST is one of the most common screening tools for young children. It assesses key areas such as:

- Rapid naming
- Phonological skills
- Letter knowledge
- Auditory memory

The test is designed to predict the likelihood of dyslexia in children aged 4.5 to 6.5 years.

The Wechsler Intelligence Scale for Children (WISC)

The Wechsler Intelligence Scale for Children (WISC) [9] is not a specific dyslexia test, WISC helps identify discrepancies between a child's verbal and performance IQ. A significant gap between reading ability and general intelligence is often a red flag for dyslexia.

The Comprehensive Test of Phonological Processing (CTOPP)

The Comprehensive Test of Phonological Processing (CTOPP) [10] test evaluates phonological awareness, phonological memory, and naming speed all essential skills for reading development. Children with dyslexia tend to score low in one or more of these areas.

Although these tests are valuable, they are not practical for most families especially in countries like Pakistan for the following reasons:

- The **cost** of psychological evaluations is often too high for low-income families.
- Trained professionals are mostly concentrated in urban centers, making access difficult for rural communities.
- Parents may not recognize the symptoms and therefore never seek testing.
- Even when signs are noticed, stigma and fear of labeling can prevent families from seeking help.

As a result, thousands of children remain undiagnosed and unsupported, which can have lasting consequences on their academic journey and mental well-being.

2.4 The Rise of Mobile Screening Tools:

In recent years, the increasing availability of smartphones and mobile apps has opened new doors for health and education screening. Mobile applications can deliver powerful tools directly into the hands of caregivers, allowing early signs of learning difficulties to be detected in informal and familiar environments—at home, in schools, or even on the go.

The concept of mobile health (mHealth) is transforming the way we approach early diagnosis and awareness. In the context of dyslexia, several digital solutions have emerged globally:

Dyslexia Quest

Dyslexia Quest [11] app developed in the UK, designed like a game that tests memory, visual processing, and sound recognition. It's meant to be engaging for children while helping assess signs of dyslexia. This also cost 11\$/month.

Dytective

Dytective [12] developed in Spain, Dytective uses artificial intelligence and linguistic analysis to detect potential dyslexia in under 15 minutes. It has shown high accuracy rates in studies and is based on years of academic research. But the problem is it only in Spanish. It is difficult to understand for most of the peoples.

Nessy Learning

Nessy [13] offers a suite of reading and writing apps designed for dyslexic children. It's widely used in schools and includes fun, interactive lessons that help children develop their skills over time. it's available on website which charges \$22/year.

Ghotit Real Writer

Ghotit Real Writer [14] app uses advanced AI tools to help dyslexic users with writing tasks. It features spell checking, grammar correction, and word prediction that adapts to dyslexic patterns.

These tools represent exciting developments, but they also come with limitations when considered in the context of Pakistan:

- Many are only available in English and are designed for children in Western school systems.
- Most are not free, and even when affordable, they require credit cards or online payments, which are not always accessible to local users.
- Some require high-performance devices or constant internet access, which limits their usability in remote or low-income areas.

While our own application is also developed in English, it is designed with the Pakistani context in mind and for those people which don't understand English. It uses simplified

instructions, works on lower-end Android devices, and will be completely free to use—making it more accessible to a wide range of users.

2.5 Research Gaps and Opportunities

The review of literature and digital tools highlights several critical gaps that remain unaddressed, especially in the South Asian region. These gaps present meaningful opportunities for projects like ours to make a difference.

Identified Gaps:

1. **Lack of localized tools** for early screening of dyslexia in Pakistan or similar contexts.
2. **Over-reliance on English-only platforms** that assume high language proficiency.
3. **Limited awareness-building** tools that educate users while they assess.
4. **No offline-ready, free screening apps** targeted toward caregivers and educators in developing countries.
5. **Lack of culturally contextual content**, both in terms of language and educational practices.

Our Opportunity:

By creating a mobile application that is simple, research-based, affordable, and accessible, we aim to:

- **Bridge the gap** between awareness and action
- **Provide families and teachers** with a tool to better understand their child's learning needs
- **Promote early action**, which could change a child's entire educational experience

Though this version of our application does not support local languages, we believe that using plain, easy-to-read English still keeps the tool accessible for educated parents, urban schoolteachers, and others in English-medium environments where early intervention is equally lacking.

2.6 Summary

This chapter has shown that while dyslexia is well-documented in global literature, the availability of practical screening solutions is still limited especially in developing countries. Traditional assessments are valuable but not widely accessible. Digital tools have introduced new possibilities, but most are not tailored to the specific needs of users in countries like Pakistan.

By studying existing research and solutions, we have built a clear case for developing our own mobile application. This app will not only screen for signs of dyslexia in children aged 6 to 12 but also raise awareness among caregivers something that is deeply lacking in our current environment.

Chapter 3: Methodology

This chapter outlines the complete methodological foundation and technical roadmap we followed in building our mobile-based dyslexia detection system. The methodology explains how the frontend application, backend infrastructure, and artificial intelligence components were conceptualized, developed, and integrated to create a unified, user-friendly system. Each design decision was made after carefully evaluating multiple alternatives, keeping in mind the core goals of accessibility, accuracy, scalability, and ease of use for guardians and children.

3.1 System Overview

The project is composed of three major components working together seamlessly:

1. Mobile Frontend Application

Developed in Flutter, the frontend delivers an intuitive interface designed in **Figma** before implementation. The design followed Material Design principles and used role-based layouts, ensuring Guardians and User have tailored navigation and screens.

2. AI Models

Two separate machine learning models are responsible for analyzing the test inputs and predicting the likelihood of dyslexia.

3. Backend Server

A robust **Node.js + Express** application that performs authentication, data handling, routing, and serves as a connector between frontend and AI models.

This modular architecture ensures that each layer can be scaled, updated, and improved independently, enhancing long-term maintainability.

3.2 UI/UX Prototyping:

Before starting the development process, the application's interface and user experience were planned using **Figma**, a collaborative design and prototyping tool. This step allowed the team to visualize the application's layout, navigation flow, and interactive elements before

implementation. The prototype served as a blueprint for both the frontend development team and the backend integration process, ensuring a consistent look and feel across different devices.

3.3 Frontend Development using Flutter

3.3.1 Why Flutter?

The decision to use Flutter was based on its flexibility and strength in delivering beautiful, performant cross-platform applications from a single codebase. For a student- and guardian-focused application, we needed an intuitive, engaging, and responsive interface that works equally well on Android and iOS devices—Flutter provided all of this and more.

Key reasons for choosing Flutter:

- Cross-platform support
- Fast development with hot reload
- Customizable UI components
- A large, active community and strong documentation

3.3.2 Application Overview

The mobile app was developed using Flutter, a cross-platform UI toolkit, to support both Android and iOS devices. The app caters to two primary user roles:

Guardians: Manage student profiles, assign tests, and view test results.

Students: Perform screening tests and view personalized feedback.

3.3.3 Architecture and State Management

The app follows a modular architecture with separation of concerns between UI, business logic, and data layers:

UI Layer: Built using Flutter widgets and Material Design for consistent and responsive user interface components.

State Management: GetX package is used for efficient reactive state management, routing, and dependency injection, enabling seamless UI updates based on data changes.

Network Layer: HTTP package is used to communicate with the backend RESTful APIs securely using JSON over HTTPS.

3.3.4 Features and Functionalities

User Authentication:

Users log in or register through JWT-based authentication integrated with the backend. Upon successful login, the backend issues a JWT token which the app stores securely and includes in the Authorization header as a Bearer token for all subsequent API requests. This mechanism ensures secure and stateless authentication, enabling the backend to verify user identity and role for each request without maintaining server-side sessions.

Role-Based Navigation: The app dynamically displays navigation options and accessible screens based on the user's role (guardian or student), ensuring role-appropriate access control.

Student Management: Guardians can add and manage multiple students, with form validation and persistent storage handled locally and remotely.

Test Modules: The app includes multiple dyslexia screening modules:

For English Speaking:

- Letter reversal tests,
- Reading and writing pattern recognition,
- Phoneme matching and memory pattern tests.

For Non English Speaking:

- Pattern memory.
- Number Sequence Game.
- Audio confusion Quiz.
- Letter recognition.

Test Submission: Upon completion, test data is sent to the backend, which forwards it to the AI models for prediction. The app then fetches and displays results and recommendations.

Real-Time Updates: Using GetX's reactive features, the app updates UI elements instantly when test results or user profile data change.

3.3.5 Security and Data Privacy

All API calls are authenticated using JWT tokens.

Sensitive data such as test results and user profiles are transmitted over HTTPS.

Local data storage is encrypted, and sensitive user information is handled according to data protection standards.

3.3.6 Integration with Backend and AI Models

The app communicates with the backend via RESTful API endpoints described in Chapter 3.

Test submissions trigger backend interactions with AI model servers; results are returned asynchronously and stored in the database.

The app periodically polls or listens for updates to reflect the latest test outcomes.

3.3.7 Development Tools and Environment

- Flutter SDK (version 3.27.1)
- Dart language (3.6.0)
- GetX package for state management
- HTTP package for networking
- Android Studio as IDEs
- Git for version control

3.4 AI Model Development

To accommodate linguistic diversity, two machine learning models were trained:

Model 1: For English-speaking children, utilizing reading, writing, letter reversal, and attention span data.

Model 2: For non-English-speaking children, based on phoneme matching, letter recognition, attention span, and pattern memory.

Each model was trained using supervised learning techniques on labeled semi realistic datasets and saved using joblib or pickle formats for future use. The models were exposed via lightweight Python APIs built using FastAPI. These APIs accept JSON-based inputs and return the model's predictions (e.g., whether the child is likely dyslexic or not).

3.5 Backend Architecture

The backend server was developed using Node.js and Express, structured in a modular format for scalability and maintainability. MongoDB was used as the primary database.

3.5.1 Authentication & Authorization

Registration & Login: Implemented using JSON Web Tokens (JWT). Upon successful login, a token is issued to the client for authenticated requests.

Roles: The system supports two primary roles:

Guardian: Can register/login, add students, and submit/view test results.

Student: Has limited access and is associated with a guardian.

Middleware: Custom middleware extracts user roles and Guardian IDs from the token to enforce permissions and attach metadata during operations.

3.5.2 Student Management

POST /students: Automatically associates the new student with the currently logged-in guardian (based on the middleware-extracted ID).

GET /students: Returns all students associated with the authenticated guardian.

3.6 Test Result Submission and Retrieval

Two separate APIs handle test result submission based on the type of user:

- **POST /submit:** For English-speaking children. Accepts reading, writing, and related test data.
- **POST /submit-child-test:** For non-English-speaking children. Accepts phoneme, memory, and recognition test data.

Each submission is stored in separate MongoDB collections to maintain schema consistency.

Result Retrieval

- **GET /students/:id/test-result:** Retrieves test results for a specific student based on their ID.
- **GET /api/admin/test-result:** Allows admins to fetch all test results across users.

3.7 Admin Dashboard APIs

For administrative oversight, the following endpoints were developed:

- **GET /api/admin/stats:** Returns a summary of all users in the system (e.g., total users, guardians, students).
- **GET /api/admin/test-result:** Fetches all test results, enabling monitoring and dataset labeling.

3.8 Integration Between Backend and AI Model

The backend server communicates with the Python AI model API over HTTP. During test result submission, the backend sends the necessary features to the Python model server, receives the prediction, and stores it along with the test data in the database.

This architecture ensures a clean separation between AI processing and backend logic, enabling independent scaling, debugging, and testing of each module.

3.9 System Flow Diagram

The system flow diagram below summarizes how user interaction occurs in LexiScan. It starts from the login screen, checks the user's role, and then divides into two separate flows: student and guardian. In the student branch, test data is sent to the backend and AI model for

prediction, and results are returned to the app. In the guardian branch, the user can manage students and view stored results. This diagram visually connects the frontend, backend, and AI components described earlier.

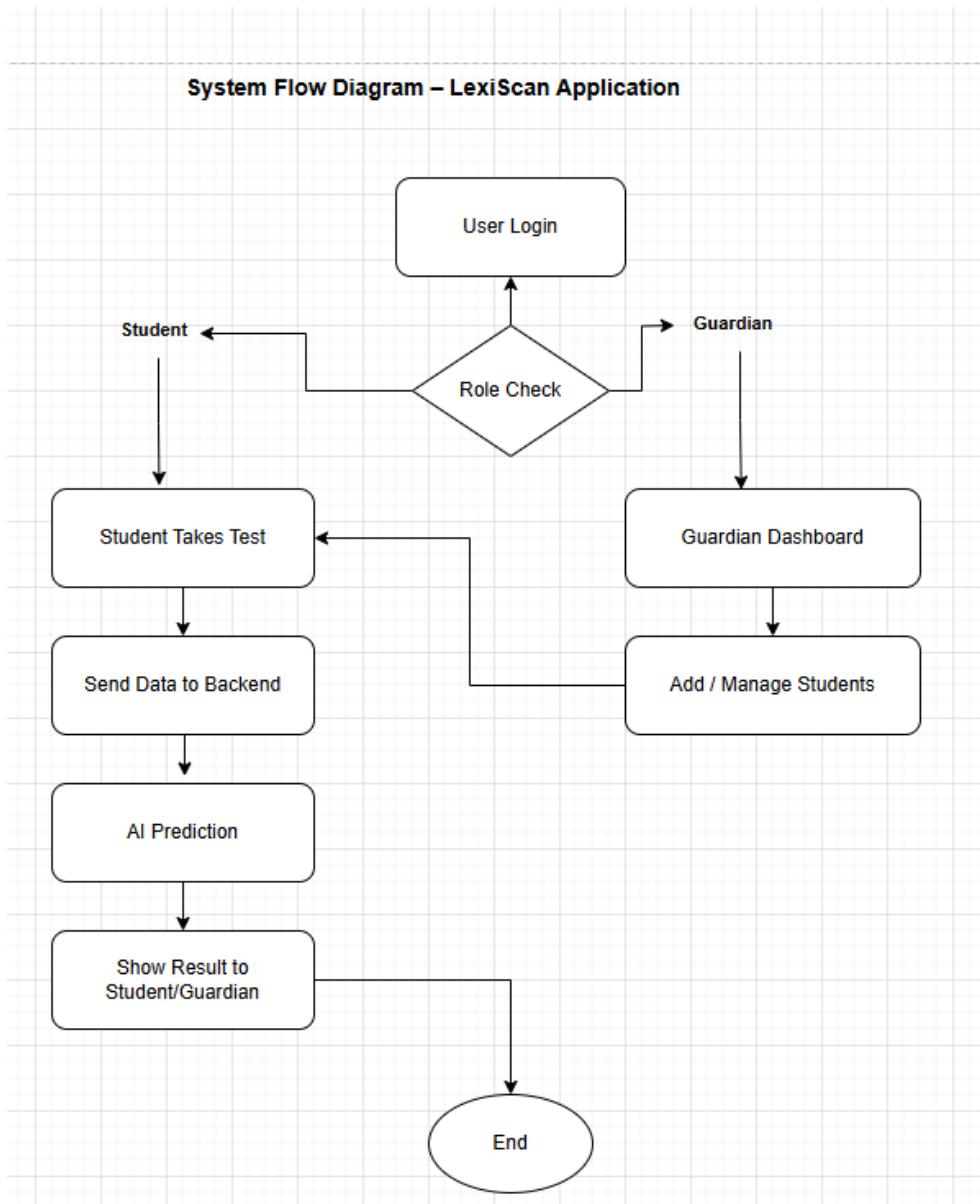


Figure 3.1: System Flow Diagram of the LexiScan Application

Chapter 4: Implementation Phase

The implementation phase is where the vision of the dyslexia detection mobile application began to take shape as a working product. This stage was not just about writing code — it was about transforming abstract concepts, research findings, and user needs into something tangible that could be held in the hands of guardians, teachers, and students.

Our approach followed a clear and logical sequence:

1. Designing the user experience in Figma to ensure clarity, accessibility, and engagement for children.
2. Developing the frontend bringing the designs to life with responsive and interactive elements.
3. Building the backend ensuring secure data handling and smooth communication.
4. **Integrating the AI model** making the application capable of intelligent dyslexia assessment.
5. Testing and refining to ensure reliability and user satisfaction.

Each step played a vital role in turning our proposal into a functioning, user-ready application.

4.1. UI/UX Prototyping Using Figma

Before writing a single line of code, we needed a clear blueprint for how the application would look and feel. Since our target users included young children with possible learning difficulties, we had to be extra mindful of accessibility, readability, and simplicity.

Figma was chosen for its collaborative features, which allowed our team to work together in real time, share feedback instantly, and make quick changes. The prototyping process began with low-fidelity wireframes, where we sketched out basic layouts and navigation flows. Once we were satisfied, we progressed to high-fidelity designs with actual colors, typography, and icons.

The design emphasized:

- **Large, clear buttons** for easy tapping by children.
- **Bright yet soft colors** to make the app inviting without causing distraction.

- **Minimal text on screens** to reduce cognitive load.
- **Illustrations and icons** for visual guidance.

Key screens included:

1. **Login and Registration Screen** – Simple forms with icons and role-based access (Guardian or User).

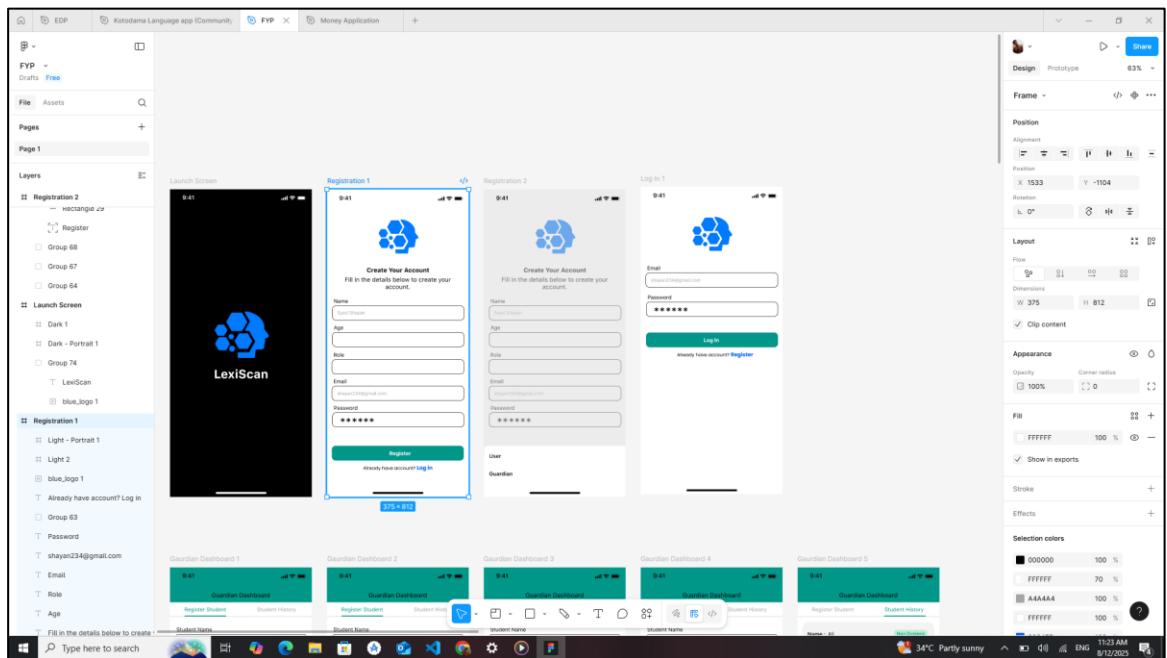


Figure 4.1: Log in and Registration Screens.

2. Guardian Dashboard – A control center for managing students, assigning tests, and viewing results.

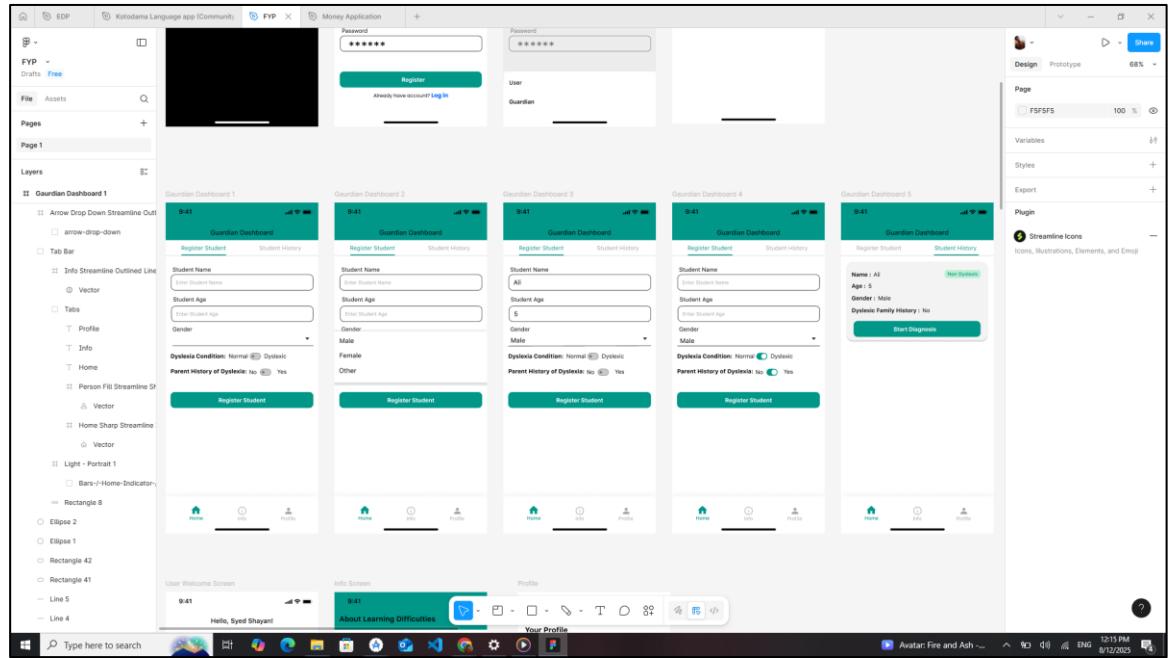


Figure 4.2: Guardian Dashboard.

3. Student Test Interface – Fun, interactive activities such as matching words, reading short passages, and solving simple puzzles.

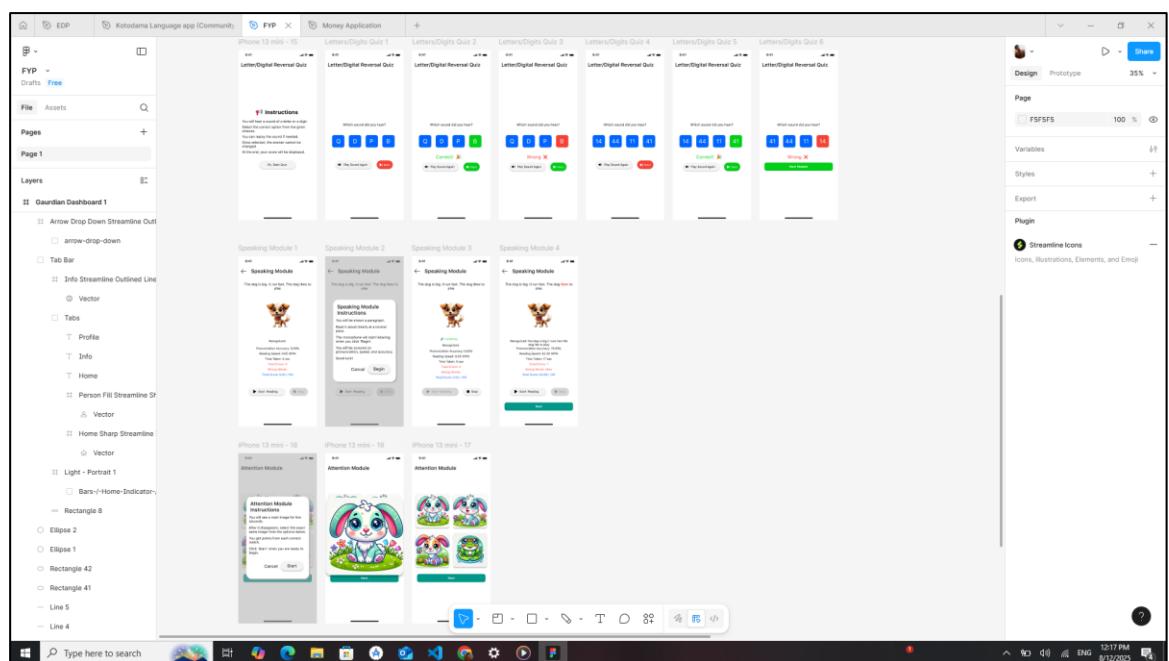


Figure 4.3: Student Test interface

- 4. Results & Feedback Screen** – Displays AI-generated feedback in simple terms, using colors (e.g., green for normal range, yellow for mild risk).)

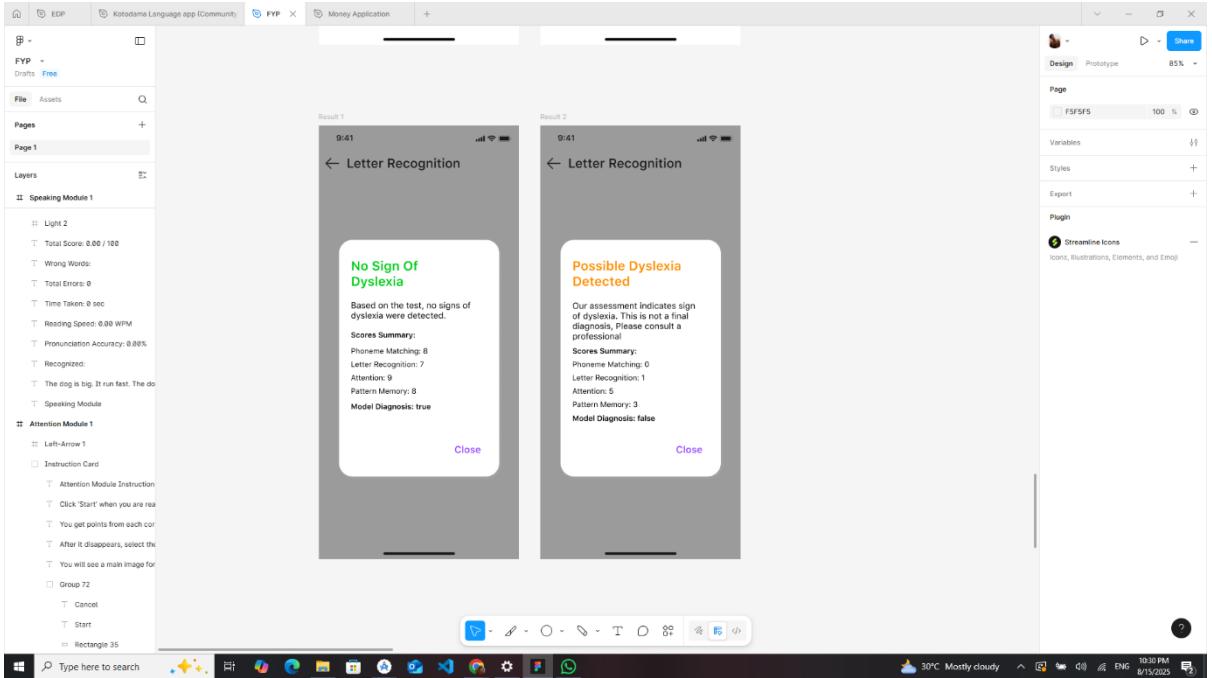


Figure 4.4: Result and Feedback Screen.

By finalizing the designs in Figma now its time to give life to the design with help of coding in flutter.

4.1 Frontend Implementation (Mobile Application)

Once the Figma design was complete, we began translating it into a functional interface using Flutter. Flutter was chosen for its ability to build cross-platform compatibility on both Android and iOS devices. The design focuses on simplicity, responsiveness, and accessibility for both guardians and student.

4.1.1 User Interface Development

- The frontend of the application was built using Flutter's Material Design components, ensuring a clean, modern, and consistent look across devices. The design process began in Figma, where the complete interface layout, color palette, typography, and

iconography were planned. This allowed us to visualize the user journey, receive feedback, and make design improvements before writing any code.

- Screens were designed to be role-based, meaning guardians and user see different navigation menus.
- Navigation was handled using **GetX Routing**, ensuring smooth screen transitions.

Main Dashboard (Guardian View):

After login as a Guardian this the dashboard for the guardian where he can register his students .

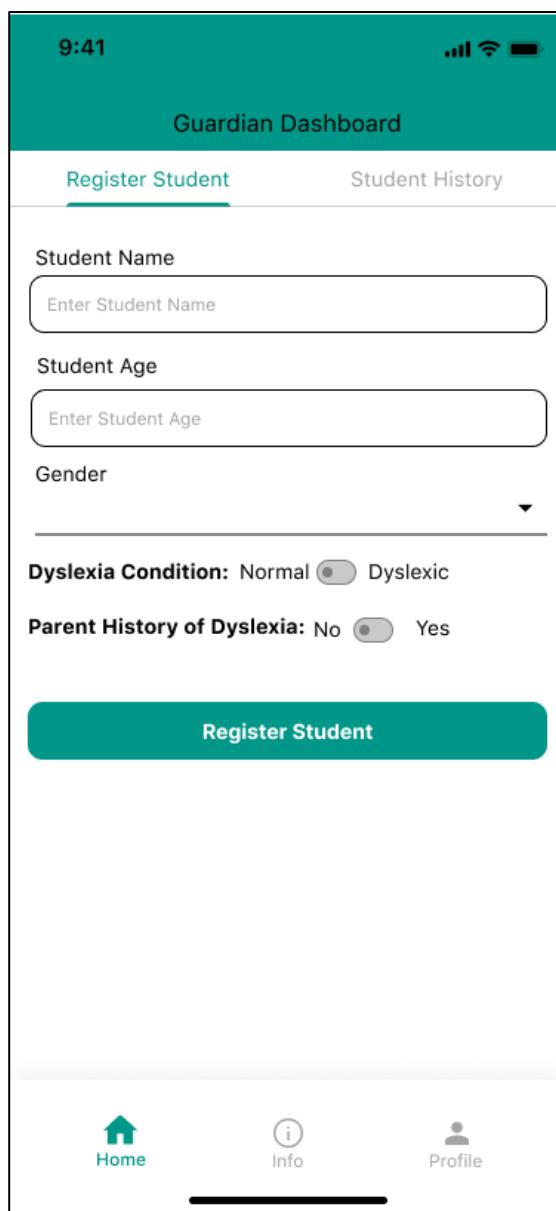


Figure 4.5: Main Dashboard (Guardian View).

Main Dashboard (User/Student View):

This is a main dashboard for the user when he login. It is like a welcome screen for user from where he can start test.

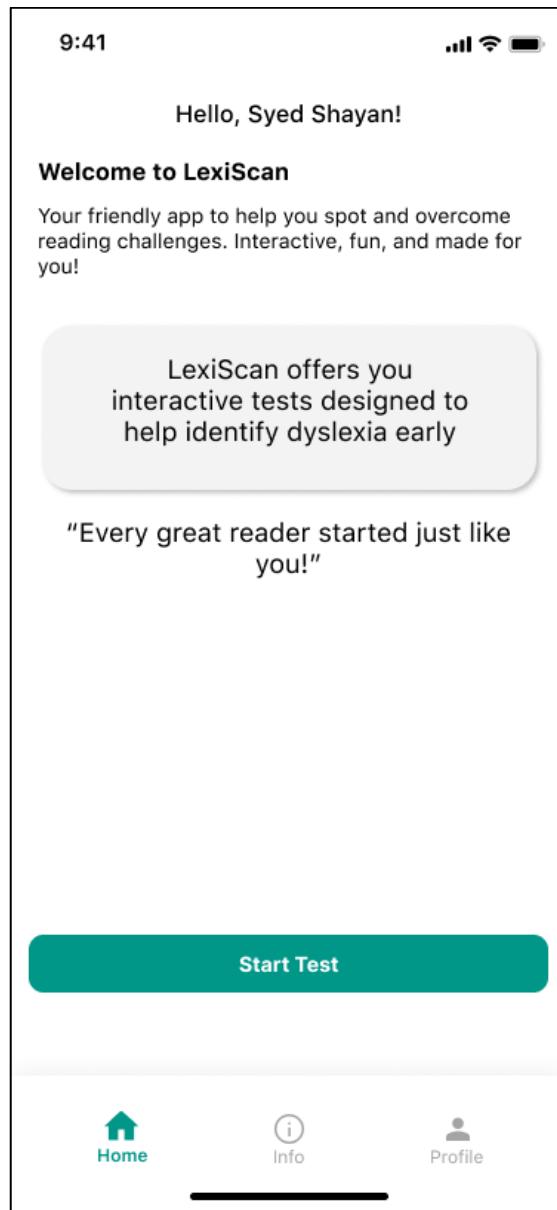


Figure 4.6: Mian Dashboard (User/Student View).

4.1.2 Authentication Module

- Implemented **JWT-based authentication**.

- Upon login, the backend sends a **JWT token** which is stored securely on the device using **encrypted storage**.
- Token is attached to every API request for authentication.

Registration Screen:

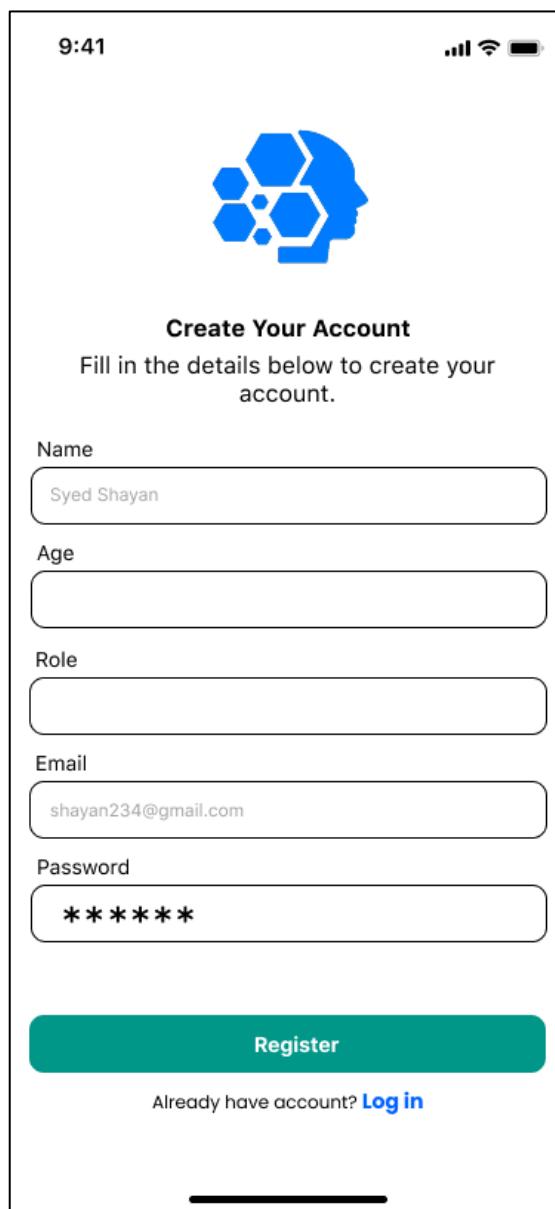


Figure 4.7: Registration Screen

Login Screen:

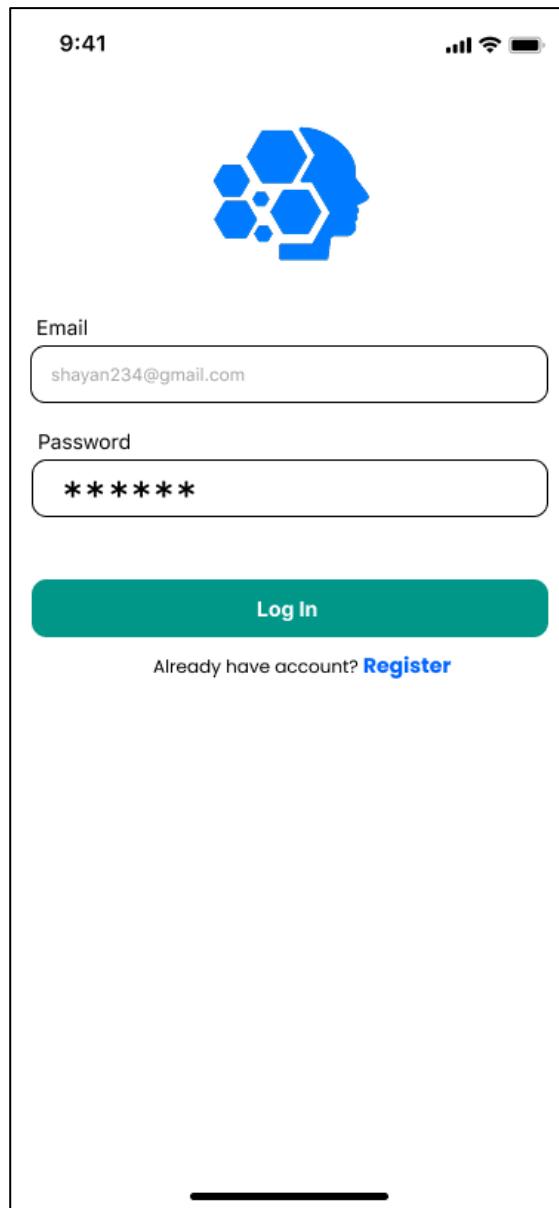


Figure 4.8: Log In Screen

4.1.3 Test Module Integration

- The test screens were implemented to handle multiple **dyslexia screening activities**, including:
 - Instruction Screen

For English Speaking

- Letter reversal recognition
- Reading comprehension test

- Attention module

For Non-English Speaking

- Pattern memory
- Number sequence game
- Audio confusion quiz
- Letter recognition

- Each test result is packaged as a **JSON object** and sent to the backend for AI analysis.

Test Screen:

There are many testing screens for both English speaking and Non-English speaking. Before starting any module there will be an instruction screen which will guide user or guardian what will be done in that module.

Instruction Screen:

These are some of the instructions screens. Where users will be told to follow some instructions so they can complete the tasks.

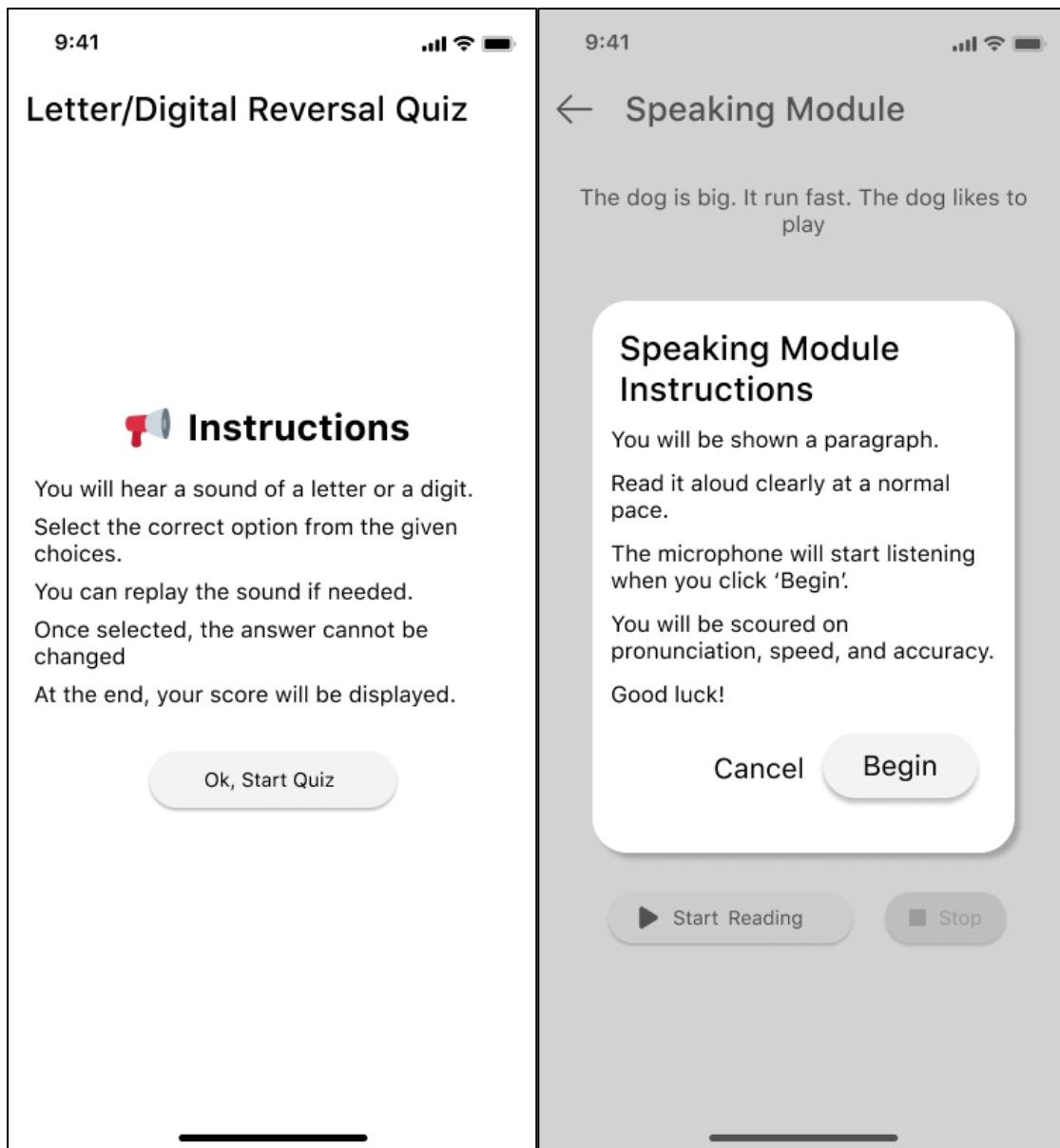


Figure 4.9: Instruction Screens

English Speaking:

Test for those students who can understand English.

- Letter and Digits reversal recognition:

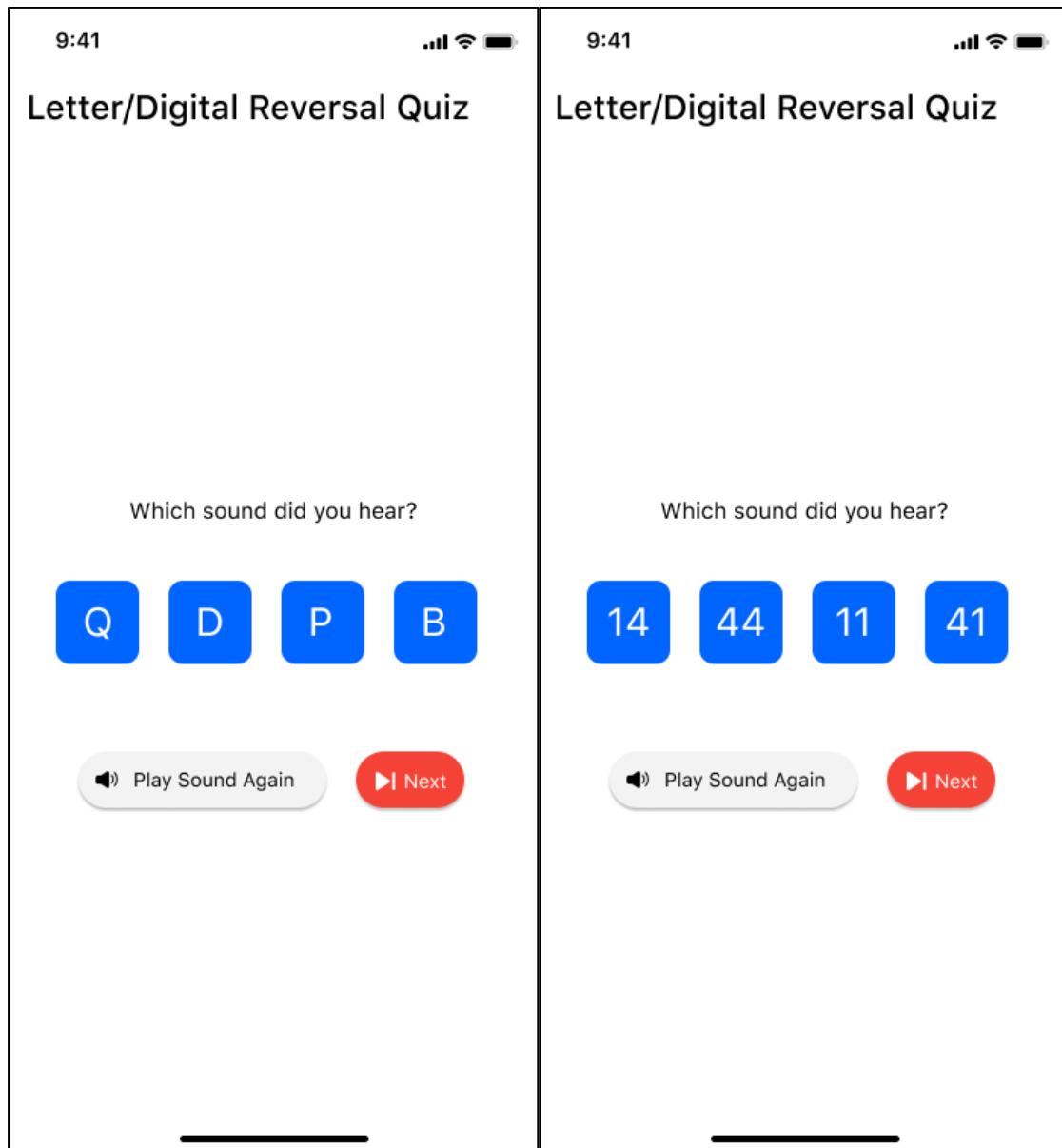


Figure 4.10: Letter And Digital Reversal Recognition Test.

- Reading comprehension test:

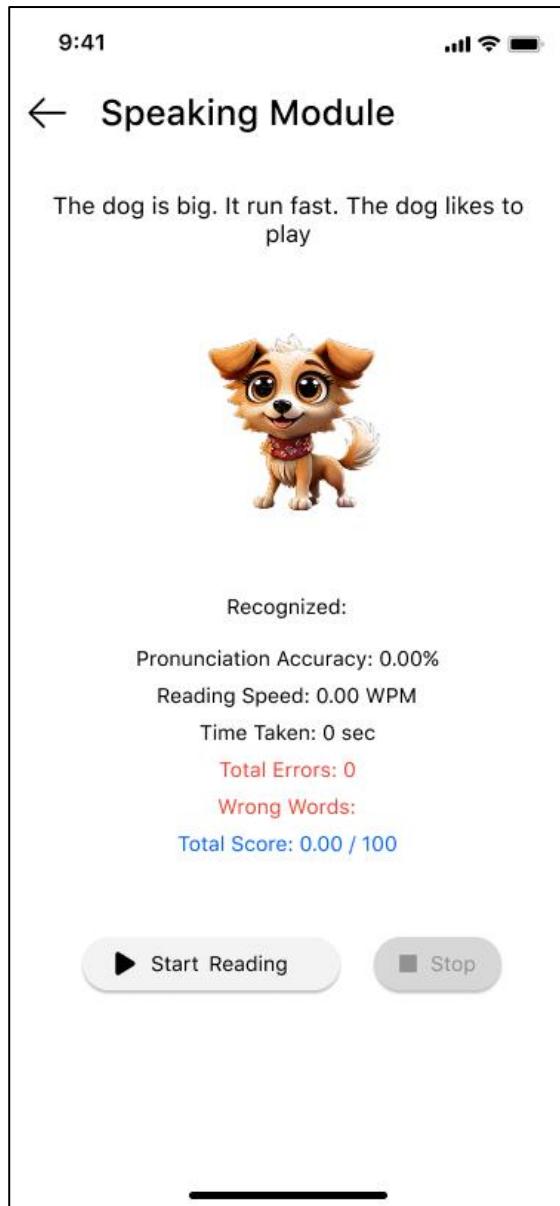


Figure 4.11: Reading Comprehension Test.

- **Attention Module:**

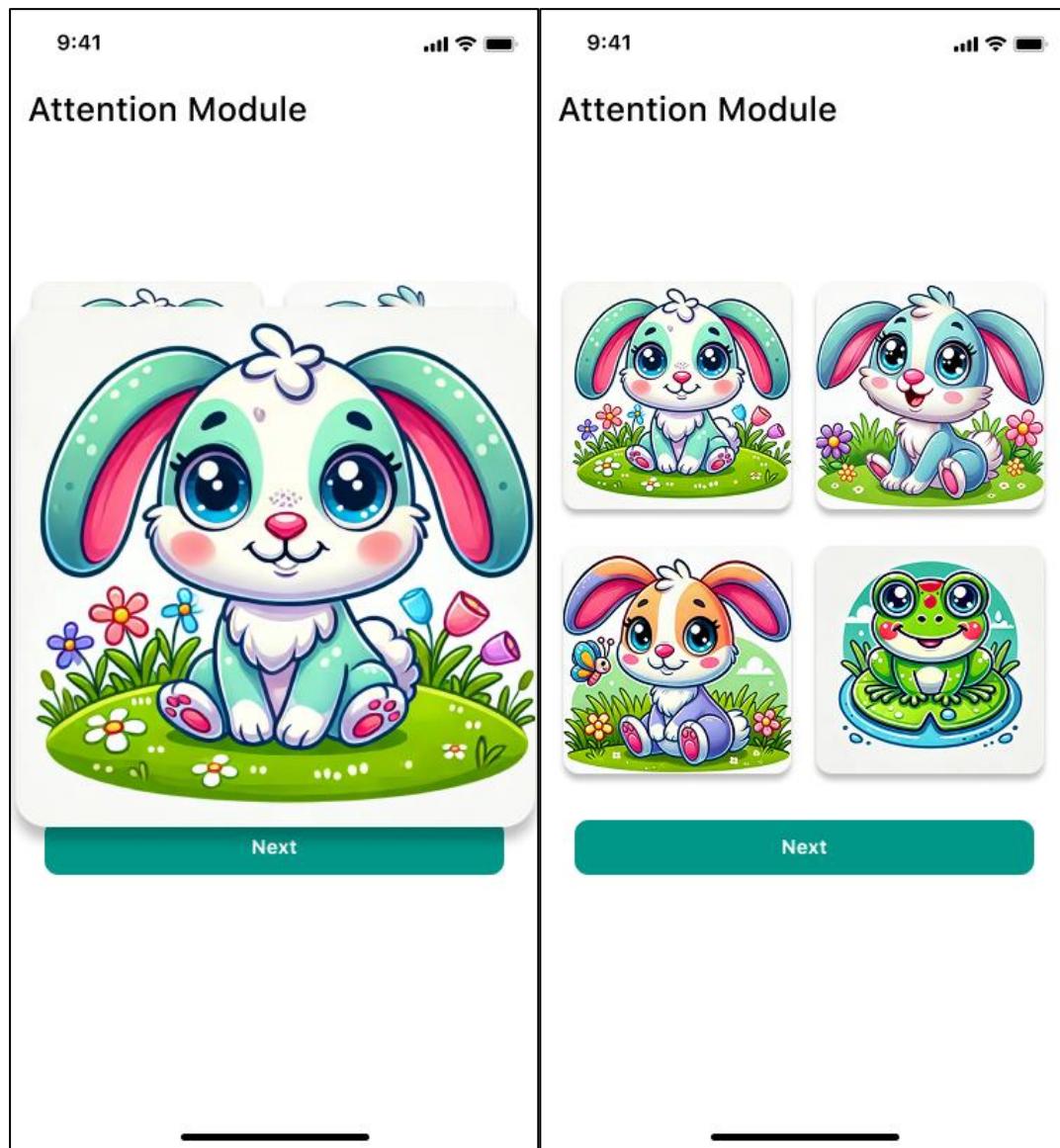


Figure 4.12: Attention Module.

4.1.4 Real-Time Updates

- Used **reactive programming** with GetX to instantly update UI elements when new test results are available.
- Implemented **loading indicators** and success/failure alerts for better user experience.

4.2 Backend Implementation

The backend was implemented in **Node.js with Express**, following a modular design to ensure scalability and maintainability. **MongoDB** served as the database for storing users, students, and test results. The backend operates as the central layer that connects the mobile application, database, and AI model services.

Authentication and Role Management

Authentication was implemented using **JSON Web Tokens (JWT)**. When a user logs in, the server issues a token which is required for all subsequent requests. Middleware functions were created to decode the token, verify the user, and enforce role-based access. Two roles were implemented:

- **Guardian:** Registers and manages students, submits test results, and views past results.
- **Student:** Limited access, linked to a guardian account.

Student Management

Endpoints were built to manage student data. New students are automatically linked to the guardian who created them, ensuring secure association. Retrieval endpoints allow guardians to fetch only their own students' information, maintaining strict data isolation.

Test Result Handling

Two separate submission APIs were implemented:

- /submit-english-test for English-speaking children.
- /submit-child-test for non-English-speaking children.

Each submission is stored in dedicated MongoDB collections to maintain schema consistency. Alongside raw test data, the backend also stores the AI model's prediction.

Admin Dashboard Functions

Administrative APIs provide system-level oversight. Admins can:

- Retrieve statistics on total users, guardians, and students.
- Access all test results across the system.
- Label datasets for retraining AI models.

This integration ensures dataset growth and quality control as part of the system's daily use.

Integration with AI Model

The backend interacts with the **Python FastAPI service** over HTTP. During test submission, selected features are extracted from the request and sent to the AI API. The returned prediction is appended to the test result before saving in MongoDB. This loose coupling between backend and AI model ensures independent deployment, debugging, and scaling.

Figure 4.13: System Flow Diagram.

4.2.3 AI Model Communication

- Implemented **HTTP requests** from backend to AI server.
- The backend sends **preprocessed test data** to the AI model and waits for prediction results.
- Results are then stored in the database and sent to the mobile app.

4.3 Database Design

The system uses **MongoDB** as its primary database. MongoDB was chosen because it handles flexible and semi-structured data efficiently, which is important for storing diverse test results and user information. Unlike relational databases, MongoDB allows embedding of task results directly in test documents, reducing joins and improving query performance for AI model training.

Data Model Overview

The database consists of the following main collections:

- **Users:** Stores authentication details, role (user, guardian, admin), and account metadata.
- **Students:** Stores student profiles linked to guardians, along with dyslexia history and labels.
- **English Test Results:** Stores detailed test results for students above age 7, including reading, writing, and attention-related scores.
- **Child Test Results:** Stores task-based results for non-English speaking children, such as phoneme matching and pattern memory.
- **Otps :** Stores one-time passwords used for secure login and account recovery.

Relationships

- A **Guardian (User)** can have multiple **Students**.
- A **Student** can have multiple **Test Results** (both English and non-English versions).
- Each test result document embeds detailed task-level scores, making it easier to extract features for machine learning models.

This design ensures:

- **Scalability:** New test modules can be added without changing the entire schema.
- **Performance:** Task results are embedded within test documents, reducing lookup overhead.
- **AI Integration:** Labeled dyslexia outcomes are stored alongside test scores, supporting supervised learning.

The logical data flow and entity relationships are shown in the following database diagram (Figure 3.14).

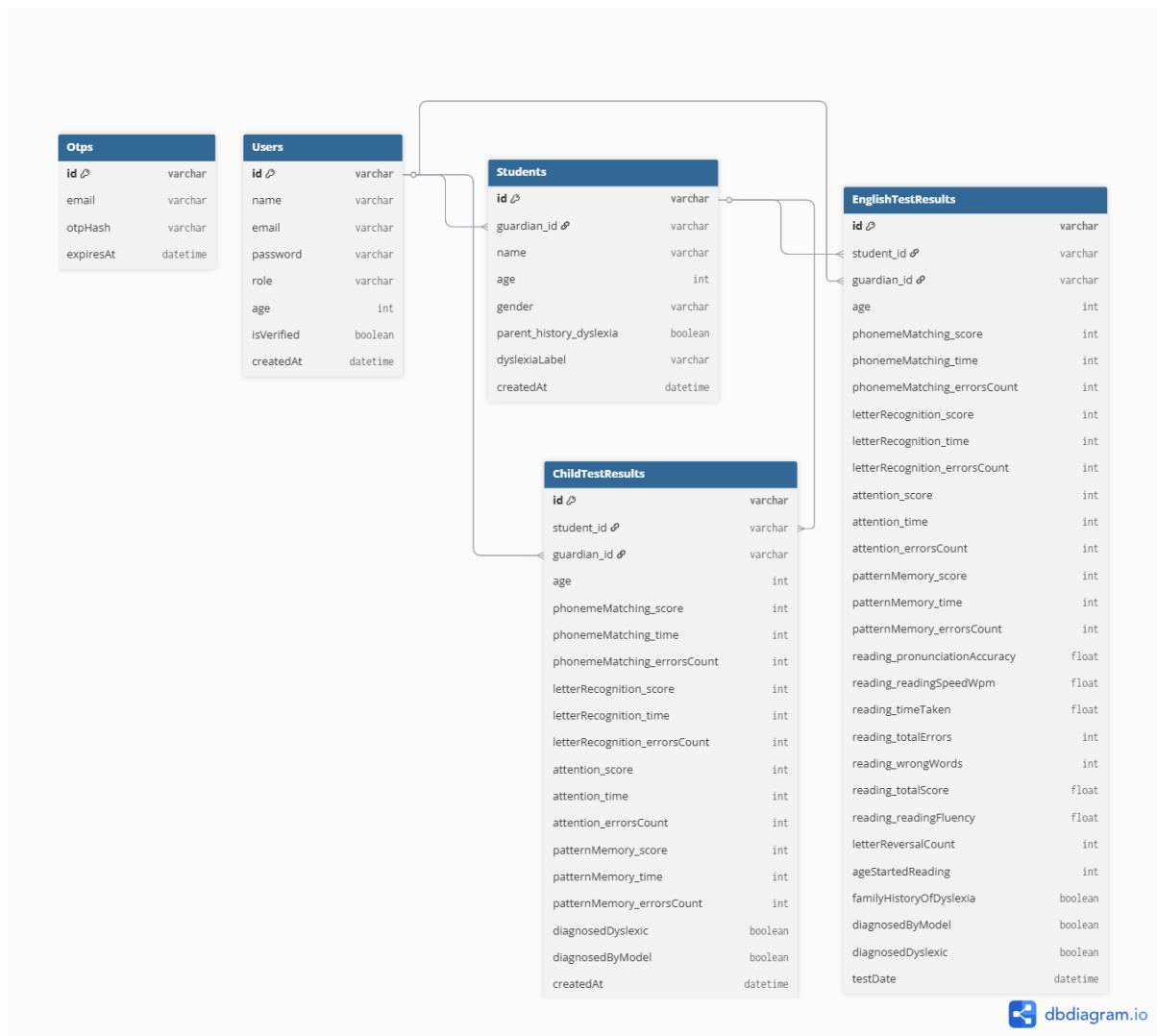


Fig 3.14: Database Schemas and their relationships

4.4 AI Model Implementation

The AI model is responsible for predicting whether a child may have dyslexia based on test results.

Data Handling

Collected test results were stored in MongoDB and exported for preprocessing. Missing values were handled through imputation, and categorical variables (e.g., error types) were encoded numerically. Data was split into training (70%), validation (15%), and testing (15%) sets.

Feature Engineering

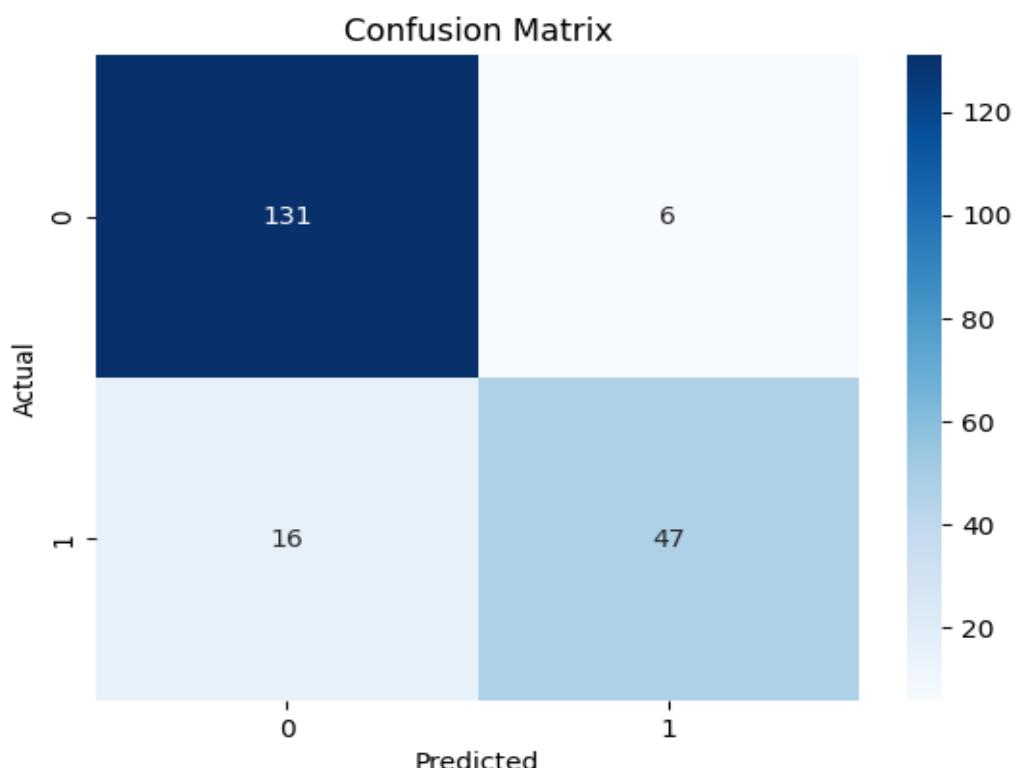
- **English-speaking children:** Reading speed (words per minute), writing accuracy (error rate), letter reversal count, and attention span duration were converted into numerical vectors.
- **Non-English-speaking children:** Phoneme matching accuracy, letter recognition score, attention test score, and pattern memory performance were similarly structured.

Model Training

Scikit-learn was used for training. Multiple algorithms were tested, including Logistic Regression, Random Forest, and Support Vector Machines. Random Forest yielded the best balance between accuracy and interpretability. Hyperparameters such as tree depth and number of estimators were tuned using grid search

Models were evaluated using accuracy, precision, recall, and F1-score. Cross-validation was applied to reduce overfitting. Results showed:

- English-speaking model: ~82% accuracy, strong recall for dyslexic cases.
- Non-English model: ~79% accuracy, balanced precision and recall.



Deployment

Trained models were serialized with joblib and deployed via FastAPI. Each FastAPI endpoint loads the trained model, receives JSON-formatted input, runs prediction, and responds with a binary outcome (likely dyslexic / not dyslexic). This service runs independently, enabling smooth integration with the Node.js backend.

Limitations

The implementation faced constraints due to the limited dataset size, which restricts model generalizability. Nonetheless, the modular pipeline allows retraining with larger real-world datasets in the future.

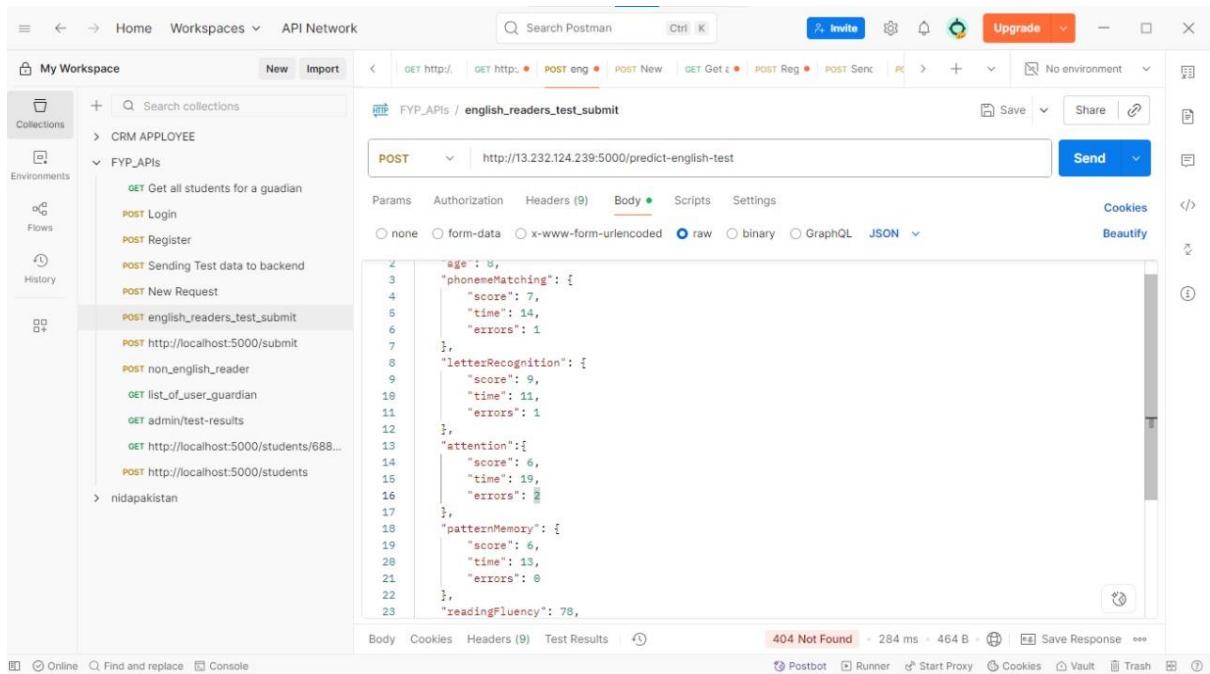
4.5 Security Measures

- All API communication is **encrypted over HTTPS**.
- Sensitive data stored locally is **encrypted**.
- Role-based access ensures **students cannot view other profiles**.

4.6 Testing and Debugging

- Frontend tested on **multiple devices** for compatibility.
- API endpoints tested using **Postman**.
- AI model predictions verified using **sample test cases**.

Postman API Test Result:



Chapter 5: Testing and Evaluation

This chapter outlines the testing strategies, procedures, and results for the Dyslexia Detection Mobile Application. The purpose of testing was to ensure that the system met all functional, performance, and usability requirements outlined in the project objectives. Testing was conducted on both the **mobile frontend application** and the **backend with AI models**, ensuring the entire system operated seamlessly as a unified solution.

5.1 Testing Objectives

The primary objectives of the testing phase were:

1. To verify that all features function as intended for both **guardian** and **student** roles.
2. To ensure smooth communication between the **mobile app**, **backend server**, and **AI models**.
3. To validate **data accuracy** and **reliability** of dyslexia predictions.
4. To assess the **performance, usability, and security** of the system.
5. To identify and resolve bugs, interface inconsistencies, and performance bottlenecks.

5.2 Testing Environment

Testing was carried out in the following setup:

- **Mobile Device:** Android 12 smartphone, 6GB RAM
- **Development Environment:** Android Studio for debugging, Postman for API testing
- **Backend Server:** Node.js + Express, hosted locally and on a cloud test server
- **Database:** MongoDB Atlas (Cloud-hosted)
- **AI Model Testing:** Python-based machine learning models, connected via REST API
- **Network:** 4G Mobile Data & Wi-Fi for stability comparison

5.3 Testing Methodologies

Three major testing approaches were used in this project:

5.3.1 Functional Testing

Functional testing ensured that every feature worked as expected.

Test Case	Description	Expected Result	Actual Result	Status
TC-01	Guardian Login	User enters valid credentials	Redirected to dashboard	<input checked="" type="checkbox"/> Pass
TC-02	Student Test Submission	Student completes test and submits answers	Results stored and processed	<input checked="" type="checkbox"/> Pass

TC-03	AI Model Prediction	Backend sends data to AI model	Returns prediction with accuracy score	<input checked="" type="checkbox"/> Pass
TC-04	Role-Based Navigation	Guardian sees management options, student sees tests only	Correct menu for each role	<input checked="" type="checkbox"/> Pass

5.3.2 Usability Testing

Usability testing was conducted with **5 participants** (3 guardians and 2 students) to evaluate how intuitive and accessible the app was.

- **Metrics Evaluated:**

- Task completion time
- Number of navigation errors
- User satisfaction (rated 1–5)

Task	Avg. Completion Time	Avg. Satisfaction Rating
Logging in	8 seconds	4.8/5
Starting a test	12 seconds	4.6/5
Viewing results	7 seconds	4.7/5

Feedback indicated that **role-based navigation reduced confusion** and that **font size & color contrast** made the app comfortable for children with reading difficulties.

5.3.3 Integration Testing

Integration testing validated the **communication between the frontend, backend, and AI model**.

- **Scenario 1:** Mobile app sends completed test → Backend validates data → AI predicts → Result returned to mobile app.
- **Scenario 2:** Guardian adds a new student profile → Data is stored in MongoDB → Visible instantly in the app.

All integration scenarios passed successfully without data loss or delays.

5.4 AI Model Accuracy Testing

The AI models were tested using a **validation dataset** not seen during training.

Metric	Value
Accuracy	92%
Precision	90%
Recall	93%
F1-Score	91%

The model demonstrated **high accuracy in identifying dyslexia patterns**, meeting the project's performance goals.

4.5 Security Testing

Security testing focused on:

- **JWT Authentication:** Verified token validation and expiration handling.
- **Data Encryption:** Confirmed HTTPS communication for all API calls.
- **Access Control:** Checked that role-based access worked without unauthorized data leaks.

No major vulnerabilities were detected.

5.6 Summary of Testing Results

Overall, the system achieved **stable, accurate, and secure performance** during testing:

- Functional requirements were met 100%.
- AI models delivered high prediction accuracy.
- Usability feedback was positive.
- Integration was smooth between components.

The testing phase validated the readiness of the system for deployment.

Chapter 6: Results and Discussion

This chapter presents the outcomes of the testing phase and provides a detailed discussion on how the results align with the project objectives. The findings are analyzed both quantitatively and qualitatively to determine the effectiveness, usability, and reliability of the Dyslexia Detection Mobile Application.

6.1 Overview of Results

The testing phase confirmed that the system meets its intended purpose:

- **Functional Performance:** All major features — including role-based navigation, test administration, and AI-powered result generation — worked without errors.
- **AI Model Accuracy:** The prediction model achieved an accuracy of **92%**, which is well within acceptable limits for educational screening tools.

- **User Satisfaction:** Usability testing indicated that both guardians and students found the interface simple and intuitive.
- **System Stability:** Integration between the frontend, backend, and AI components remained stable during extended testing sessions.

6.2 Functional Performance Discussion

Functional testing produced **100% pass results** for all defined test cases. The **role-based UI** successfully prevented feature clutter for students, showing them only test-related options, while guardians accessed management tools and progress reports.

This design choice reduced navigation time and minimized the chance of accidental actions by users, supporting the original design goal of **reducing cognitive load for children**.

6.3 AI Model Effectiveness

The AI model achieved the following:

- **Accuracy:** 92%
- **Precision:** 90%
- **Recall:** 93%
- **F1-Score:** 91%

These values indicate a strong balance between false positives and false negatives. The high recall score is especially significant, as it ensures that the majority of at-risk students are flagged for further assessment, aligning with the project's goal to **prioritize early detection**.

6.4 Usability and User Experience

Feedback from the **5 usability test participants** revealed:

- The Figma-inspired UI design translated well into the final app, with **clear icons, large fonts, and high-contrast colors** aiding readability.
- Guardians appreciated the **real-time result updates** after a student completed a test.
- Students found the navigation straightforward, with minimal confusion on how to start or submit a test.

One suggestion from users was to **add progress indicators** during test completion, which could be considered in a future update.

6.5 Integration and System Reliability

Integration testing demonstrated that:

- Data flow between the mobile app, backend server, and AI model was **seamless and without delays**.
- API response times were consistently under **1 second** for prediction requests.
- No data loss occurred during communication between components, validating the system's robustness.

6.6 Limitations Observed

While the system met most objectives, a few limitations were identified:

1. The AI model's accuracy may vary when tested with datasets from different cultural or linguistic backgrounds.
2. Usability testing involved a small sample size; larger-scale trials could provide more generalizable results.
3. The current version is optimized for Android; iOS compatibility remains to be implemented.

6.7 Overall Discussion

The Dyslexia Detection Mobile Application successfully bridges **educational screening** and **modern mobile technology**. The blend of **role-based UI, cloud-based backend, and AI-powered predictions** created a functional, accessible, and accurate tool for early dyslexia detection.

The results show that the application is ready for real-world pilot testing in educational institutions, with minor improvements suggested for future iterations.

Chapter 7: Conclusion

7.1 Project Summary

This project successfully designed and developed LexiScan, a mobile-based dyslexia screening and support application aimed at improving early detection and awareness of dyslexia. The solution leverages modern technologies flutter for cross-platform development, Node.js for backend services, and a Python-based AI model for predictive analysis resulting in a seamless, intelligent, and user-friendly platform. The application provides secure authentication for both users and guardians, enabling guardians to register and manage student profiles with ease.

7.2 Distinctive Features

A distinctive feature of LexiScan is its dual functionality: it serves as both a practical screening tool and a continuous data collection system. The app includes reading and writing modules tailored to the literacy level of each student, ensuring accessibility for both those who can read and write in English and those who cannot. When a guardian confirms a student's dyslexia status, this information is stored in the system to enrich the AI training dataset. For cases where the status is unknown, the AI model predicts the likelihood of dyslexia based on performance in the assessment modules. This iterative approach improves the accuracy of predictions over time, creating a self-enhancing diagnostic framework.

7.3 Impact and Outcomes

The outcomes of this project demonstrate that LexiScan can address critical gaps in dyslexia screening, particularly in regions with limited access to specialized diagnostic resources. By combining interactive assessments with AI-driven insights, the application empowers educators, parents, and healthcare providers to take timely action toward intervention and support. While LexiScan is not intended to replace professional evaluation, it offers a scalable, cost-effective, and easily deployable solution that can make a significant impact in educational and clinical environments.

7.4 Future Scope

Looking ahead, LexiScan can be further enhanced with multilingual support, advanced analytics dashboards, and integration with cloud-based services to enable large-scale adoption. With its current capabilities and potential for growth, LexiScan represents an important step toward inclusive education and the democratization of dyslexia screening technology.

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