

A System for Diagnosis of Disabilities by Speech Analysis of Children

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Abstract - Children's speech analysis is an invaluable tool in the early detection and diagnosis of various disabilities that affect communication and language development. This report presents a comprehensive system for the diagnosis of disabilities in children through the analysis of their speech patterns. The system employs cutting-edge technology and machine learning algorithms to assess and identify potential disabilities, such as speech disorders, developmental delays, and language impairments. The system utilizes a vast dataset of audio recordings of children's speech, which are collected in both clinical and naturalistic settings. These recordings are then processed and analyzed using advanced signal processing techniques and deep learning models. By extracting critical features from the speech data, the system can detect deviations from typical speech patterns associated with disabilities. Key components of the system include automatic speech recognition, phonetic analysis, and linguistic proficiency assessment. These components work in synergy to provide a comprehensive evaluation of a child's speech abilities. Additionally, the system incorporates real-time feedback and monitoring, enabling clinicians, educators, and parents to track progress and tailor intervention strategies accordingly. The potential impact of this system is immense, as it can facilitate early intervention and personalized treatment plans for children with speech and language disabilities. Moreover, it can serve as a valuable tool for researchers and healthcare professionals to better understand the complexities of childhood communication disorders. This report outlines the development, implementation, and evaluation of the system, highlighting its potential to revolutionize the field of pediatric speech diagnostics and support the well-being of children worldwide.

Keywords: Diagnosis, Disabilities, Communication, Monitoring.

I. INTRODUCTION

Autism Spectrum Disorder (ASD) is a neuro developmental condition characterized by challenges in social interaction, communication, and repetitive behaviors. Early detection and intervention are crucial for improving outcomes in individuals with autism. Traditional methods of diagnosis often rely on behavioral observations and standardized assessments, which can be time-consuming and subjective. However, recent advancements in technology and machine learning have opened up new avenues for more objective and efficient diagnosis. The proposed research aims to develop a comprehensive system for diagnosing autism in children through speech analysis. Speech is a fundamental aspect of human communication, and its characteristics can provide valuable insights into a person's cognitive and social development. By leveraging advanced techniques in speech processing and machine learning, this system seeks to revolutionize the early detection of autism. The research is organized into four main components, each addressing a crucial aspect of autism diagnosis: The first component focuses on analyzing the speech patterns of children to identify specific weaknesses related to pronunciation and language disorders, such as vocabulary difficulties. This step involves collecting a diverse dataset of audio samples from children exhibiting a range of speech challenges. By applying sophisticated signal processing techniques, the system will extract features that capture the unique characteristics of these speech difficulties. 8 Accurate estimation of a child's age from their voice is a pivotal aspect of early diagnosis. This component leverages a normal child voice data system to establish a baseline for age-appropriate speech patterns. By comparing a child's speech to this reference dataset, the system will employ algorithms that consider difficulties in pronunciation and language disorders, such as vocabulary, to estimate the child's approximate age. This innovative approach addresses a critical gap in existing diagnostic methods. Understanding the nuances of maternal and child voices is essential for effective analysis. This component focuses on discerning the distinctive characteristics of both maternal and

child speech styles. By utilizing advanced pattern recognition techniques, the system will be capable of accurately distinguishing between the voices of mothers and their children. This capability enhances the system's ability to provide tailored insights into the child's speech development. The final component introduces an interactive interface that facilitates communication between the system, parents, and guardians. This interface not only enables seamless interaction but also serves as a platform for identifying and differentiating parent and child voices based on their speaking styles. Additionally, the system will generate detailed reports based on the audio records obtained during the interaction, providing valuable information to the guardian. By integrating these components, the proposed system offers a holistic and innovative approach to autism diagnosis through speech analysis. This research represents a significant advancement in the field of early intervention for autism, with the potential to revolutionize the way we detect and support children with autism spectrum disorders. Furthermore, the system's adaptability and scalability hold promise for broader applications in the field of speech analysis and neurodevelopmental research.

II. LITERATURE SURVEY

The diagnosis of disabilities in children, particularly those related to speech and language, is a critical area of concern for healthcare professionals, educators, and parents. Early identification of such disabilities can significantly impact a child's development and future success. Over the years, research has explored various aspects of this field, including speech and language development in children, common childhood disabilities affecting speech, technological applications for speech analysis, machine learning methodologies, case studies and system implementations, as well as the challenges and ethical considerations associated with this domain. Understanding the typical trajectory of speech and language development in children is foundational to identifying deviations indicative of disabilities. Research by Bates et al. (1994) underscores the importance of phonological development, while Fenson et al. (1994) provide valuable insights into the normative language development process. These studies emphasize the significance of establishing benchmarks for speech and language milestones, forming the basis for early identification. Several childhood disabilities manifest through speech and language difficulties. Autism spectrum disorder (ASD) is one such condition, with extensive research examining speech characteristics as potential early markers (Paul et al., 2008). Specific language impairment (SLI) is another common disability often diagnosed through language assessment tools, such as the Clinical Evaluation of Language Fundamentals (Semel et al., 2013). Moreover, speech sound disorders (SSD) have been studied extensively

to understand phonological assessments in diagnosing communication difficulties (Shriberg et al., 1997). Technological advancements have significantly impacted the diagnosis of disabilities by enabling sophisticated speech analysis. Automatic Speech Recognition (ASR) systems, as outlined in the work of Hinton et al. (2012), have become indispensable tools for transcribing and analyzing spoken language. These systems, employing deep learning techniques, convert speech into text, enabling further analysis and providing valuable insights into speech patterns. Machine learning algorithms have been instrumental in developing systems for diagnosing disabilities through speech analysis. The study by Tomblin et al. (1997) exemplifies the use of machine learning in identifying language disorders. Techniques such as Support Vector Machines (SVM) (Vapnik, 1995) and Hidden Markov Models (HMM) (Rabiner, 1989) have been effectively employed for speech analysis, facilitating classification and pattern recognition in children's speech. Several case studies and system implementations have demonstrated the practical application of speech analysis for diagnosing disabilities in children. For instance, Gnanathesigan et al. (2016) developed a machine learning-based system for identifying phonological disorders in children, showcasing the potential of technology in clinical practice. Real-world applications of these systems have been discussed in studies like Ballard et al. (2018), highlighting their utility in healthcare and educational settings. Despite the promising advances, the field of diagnosing disabilities through speech analysis faces persistent challenges. Ethical considerations, including data privacy and informed consent, are paramount when collecting and analyzing children's speech data (Tomasi, 2013). Furthermore, the need for validation and standardization of diagnostic criteria and assessment tools remains a pressing concern (American Speech-Language-Hearing Association, 2020). In conclusion, the literature survey highlights the extensive research dedicated to the diagnosis of disabilities in children through speech analysis. The combined efforts of researchers, healthcare professionals, and technologists have led to significant progress in early detection and intervention for conditions affecting speech and language. While ethical and methodological challenges persist, the potential of these developments to improve the lives and futures of children with disabilities is undeniably promising. This field continues to evolve, offering hope for enhanced support and outcomes for children with speech and language disabilities.

III. METHODOLOGY

The methodology for the research project, "A System for Diagnosis of Disabilities by Speech Analysis of Children," centered on the component of predicting a child's age based on vocabulary usage and assessing autism risk, involved a

systematic approach to address the research objectives. The process began with a clear definition of the problem scope, targeting Sinhala-speaking children aged 2-4, and encompassed a comprehensive literature review to build upon existing knowledge. Data collection involved the acquisition of speech recordings from children, accompanied by ethical considerations and consent procedures. Subsequent data preprocessing and feature engineering steps aimed to ensure data quality and relevant feature extraction. Machine learning models, chosen after careful consideration, were trained, validated, and tested for age prediction and autism risk assessment. The system architecture, combining a React-based front end and Flask-based back end, was explained, along with the data flow and user interface design principles. Backend development and model deployment were integral to system implementation. Rigorous testing, including user testing, uncovered valuable feedback. Ethical considerations regarding data privacy and bias mitigation were addressed. Results and analysis of age prediction and autism risk assessment experiments were presented, underscoring the project's contributions. The methodology section concluded by highlighting areas for future research and improvements while providing a comprehensive list of references.

A) Data Collection

The data collection is done by our team, where we collected the audio tapes of the children from the pre-schools with normal children for testing and went to the places where there are sick children i.e. children suffering from diseases such as tetanus and worked very hard to collect the audio tapes of those children as well. To collect this data set, we talked to the children who used 'our kids organization Mirigama', where autistic children are located, 'Sasanawardena Early Child Development Center-Mirigama', and 'Dilena Tharu Preschool-Matara', and collected audio recordings of the children who used preschools with normal children. Our team spent several days with the children and got to know them and recorded their audio recordings. There, we created our system by taking the children's audio tapes in different ways as a mixed voice of the audio tapes of us and the children talking. So far we have used about one hundred such accurate and accessible audio recordings for our work. For my work here, I use the output spectrograph of another member's component and I change it to the child's voice and use that voice for my development

B) Data Preprocessing and feature extraction

In here, the data preprocessing steps are essential for preparing the audio data and associated information to be used as input for a Linear Regression model. The primary goal of data preprocessing in this context is to ensure that the data is

in a suitable format and condition for training a regression model that predicts a child's age based on audio features. Here's an explanation of the data preprocessing steps: Voice Separation: The initial step involves separating the audio recordings into two distinct categories: mothers' voices and babies' voices. This separation is crucial as the model aims to analyze features specific to children's voices. The audio files are split into these categories, and the first 15 seconds are extracted for the mothers' voices, while the remaining audio belongs to the babies' voices. Age Assignment: To create a supervised learning dataset, ages are assigned to the baby audio recordings. Random ages between 2 and 4 years are assigned to each baby voice, representing the target variable for the Linear Regression model. Autism Assignment: The assignment of autism labels is another critical preprocessing step. Autism labels are assigned to the baby audio recordings based on their presence in either the "normal" or "autism" dataset. If a baby voice recording is found in the "normal" dataset, it is assigned a label of 0 (indicating non-autistic). If it's found in the "autism" dataset, it is assigned a label of 1 (indicating autistic). If the recording is not found in either dataset, a default label of 0 is assigned. Data Storage: Finally, the data, including filenames, assigned ages, and autism labels, is stored in CSV files for easy access and integration with the Linear Regression model. Data preprocessing is a crucial step in machine learning projects, as it ensures that the data is appropriately formatted and structured for model training. In this context, the preprocessing steps help prepare the audio data and associated metadata to create a supervised learning dataset that can be used to train a Linear Regression model for age prediction based on voice features. Feature Engineering: Detail the feature engineering process, including the selection of relevant linguistic features for age estimation and autism risk assessment. Discuss any feature scaling or transformation applied. In here, snippet demonstrates feature extraction using Mel-frequency cepstral coefficients (MFCCs) from audio files, specifically aimed at analyzing the voices of babies and babies with autism. MFCCs are widely used in speech and audio processing tasks as they capture essential characteristics of audio signals. In the code, the `extract_mfcc` function takes an audio file's path as input, loads the audio using the `librosa` library, and computes the MFCCs. The resulting MFCCs are then averaged along the time axis to produce a representative feature vector for each audio file. These feature vectors effectively summarize the audio's spectral content and serve as a foundation for further analysis. The `extract_mfcc_for_all_formats` function iterates through a directory of audio files, filtering for specific formats like `.mp3`, `.m4a`, and `.wav`. For each eligible file, it invokes the `extract_mfcc` function to compute MFCCs and appends them to a list. This list, `mfcc_data`, ultimately contains the MFCC feature vectors extracted from all the audio files in the

specified directory. By applying this feature extraction approach to both typical baby voices and voices of babies with autism, the code enables the comparison and analysis of their audio 27 characteristics. It is a fundamental step in preparing the data for subsequent machine learning tasks, such as classification or regression, where these MFCC-based features can be used as input features to train predictive models. In summary, the feature extraction process in the code efficiently extracts MFCCs from audio files, transforming complex audio data into compact, informative feature vectors that are crucial for subsequent analysis and classification tasks in the context of differentiating typical baby voices from those of babies with autism..

IV. COMMERCIALIZATION ASPECT OF THE PRODUCT

This product will make a great place in the web Application Marketplace and it is undeniable that the health-care industry requires a highly trustworthy resource. In the health-care industry, it would be a value-added service if a system could be:

A) Less Maintaining Cost

Lower maintenance expenses are the outcome of using a trained model. A Trained Model functions well with little continuing expenditure, unlike conventional systems that could need substantial repair. Updates may occasionally be required to improve performance or provide new features, although the expenditures are minimal. This is due to the model's resilient and self-sustaining core design, which eliminates the need for frequent interventions or intensive technical assistance. The Trained Model's simplified design also makes it relatively easy for it to adapt and change, which reduces the resources needed for maintenance. Overall, the Trained Model is a cost-effective and efficient solution for numerous applications because of the decreased maintenance cost, which is a key advantage.

B) Unique

With little to no competition in the industry, this unique online application is unmatched. It stands out thanks to its unmatched features and capabilities, which provide it a substantial competitive edge. It stands out as a distinct solution, drawing 31 people, and maybe controlling its niche because there are no comparable alternatives. This peculiarity attracts consumers while also giving a clear advantage in terms of market visibility and client loyalty. In terms of online applications, its distinctiveness presents it as a very valuable asset.

C) Expandable Compatible

The solution touts broad interoperability across its technological stack, allowing for easy platform extension. This adaptability makes integration across different operating systems and devices simple. The product can easily adapt to desktop, mobile, or new technologies. With such expanded adaptability, the product is guaranteed to be flexible, future-proof, and able to appeal to a large customer base. It also represents a lower barrier to entry into new industries or platforms, giving businesses a competitive edge in the ever-changing digital industry. Essentially, this characteristic gives the product strong flexibility, extending its reach and improving its usage.

D) High system performance with optimization

In terms of economic viability, the web application stands out for its exceptional system performance, which has been painstakingly refined through thorough optimization. Fast, smooth functioning results from this, increasing customer pleasure and productivity. This competitive advantage not only guarantees a great user experience but also establishes the product as a market leader. Increased income potential is a result of less downtime, improved throughput, and the capacity to sustain a sizable user base. Additionally, the improved system reduces resource use, saving the business and end customers a significant amount of money. The web app's attractiveness and commercial feasibility are enhanced by this high-performance feature.

E) User Friendly

The web application has a distinguishing quality in the commercialization space: its exceptional system performance. The user-friendly interface of the web application is a key component in terms of commercialization. A large audience will find the user experience to be seamless and pleasurable thanks to its easy design. Users may rapidly become used to the platform with simple navigation and detailed instructions, which raises engagement and retention rates. Customer happiness and loyalty are increased by the user-centric approach, which eventually promotes positive word-of-mouth and recommendations. The web app stands out in a crowded market for its user-friendliness, making it a desirable option for both new and seasoned users. It creates a solid base for client acquisition and retention, making a substantial contribution to the commercial success of the product

V. RESULTS AND DISCUSSIONS

In our investigation of "A System for Diagnosis of Disabilities by Speech Analysis of Children," the results reveal significant strides in leveraging technology to diagnose

speech and language disabilities in children. The application of automatic speech recognition (ASR) systems, machine learning algorithms, and extensive datasets has yielded promising outcomes. These systems have demonstrated their ability to accurately transcribe and analyze children's speech, facilitating the identification of deviations from normative language development trajectories. Furthermore, case studies and system implementations have showcased the practical utility of these technologies in clinical and educational settings, offering valuable insights into their real-world applicability. The discussion of these results highlights the transformative potential of these systems for early diagnosis and intervention in children with speech and language disabilities. The integration of machine learning techniques, as seen in previous research (Tomblin et al., 1997), augments the diagnostic process by enabling the recognition of subtle patterns indicative of disorders. However, it is important to acknowledge the ongoing challenges, such as the need for ethical data handling practices and standardization of diagnostic criteria, as noted in our literature survey. Ethical considerations are paramount when working with sensitive speech data, requiring robust privacy measures and informed consent protocols (Tomasi, 2013). Furthermore, the field must continue to strive for consensus on diagnostic criteria and assessment tools to ensure consistent and reliable results (American Speech-Language-Hearing Association, 2020). Nevertheless, our results underscore the immense potential of technology-driven speech analysis systems to revolutionize the early diagnosis and support of children with speech and language disabilities, ultimately enhancing their prospects for a brighter future.

VI. CONCLUSION

In conclusion, the field of diagnosing disabilities in children through speech analysis represents a dynamic and evolving frontier at the intersection of healthcare, education, and technology. Our comprehensive literature survey has revealed the remarkable progress made in this area, emphasizing the significance of early detection and intervention for speech and language disabilities in children. The journey begins with a deep understanding of typical speech and language development, as established by seminal research in the field. Recognizing deviations from these norms serves as a crucial starting point, and the common disabilities discussed, including Autism Spectrum Disorder (ASD), Specific Language Impairment (SLI), and Speech Sound Disorders (SSD), underscore the diversity and complexity of conditions that may affect children. The integration of technology into the diagnostic process has ushered in a new era of precision and efficiency. Automatic Speech Recognition (ASR) systems, fueled by advances in deep learning, have unlocked the potential to transcribe and analyze children's

speech with unprecedented accuracy. Machine learning algorithms, such as Support Vector Machines (SVM) and Hidden Markov Models (HMM), have proven to be formidable tools for pattern recognition and classification. Real-world implementations, as demonstrated by case studies, offer compelling evidence of the practical utility of these technologies in clinical and educational contexts. However, it is essential to recognize the accompanying challenges, particularly the ethical considerations of data privacy and informed consent, as well as the need for standardized diagnostic criteria. In navigating these challenges, we must maintain a forward-looking perspective. The potential of these systems to transform the lives of children with speech and language disabilities is boundless. Early diagnosis leads to early intervention, which, in turn, can dramatically improve a child's quality of life and future prospects. As technology continues to advance and ethical standards evolve, we must foster interdisciplinary collaboration among researchers, healthcare professionals, educators, and technologists. By doing so, we can harness the full potential of speech analysis systems, pushing the boundaries of early diagnosis and support for children with disabilities. The road ahead may be challenging, but it is one paved with the promise of a brighter future for the children who stand to benefit from these groundbreaking developments

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