

## App Based Digital Audiometer with Different Frequencies

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ARTICLE INFO	ABSTRACT
Received: 11 Dec 2024	<p>Our project aims to innovate an app-based digital audiometer with different frequencies that can be used to provide accurate results of hearing tests among all age groups. Our Mobile App conducts hearing tests using different audio frequencies for each age group. Smartphone Technology provides accessibility and Flexibility which reduce the need for specialized equipment and medical settings. It uses adaptive algorithms to adjust the parameters based on the user's age and ensure precise results. Our solution aims to early detection, and managing hearing loss, more accessible and reliable. This innovative app utilizes the widespread use of smartphones and wearable devices which allows for a portable and accessible method of hearing assessment. This application integrates reliable calibration processes to give precise and consistent results. Apart from this our application is more compatible with Android platforms ensuring its availability for users in worldwide.</p> <p><b>Keywords:</b> Hearing Loss Detection, Age wise Hearing Detection, Detecting Early Hearing Loss, Deafness Detection, Detect Hearing Problem.</p>
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### I. INTRODUCTION

Hearing loss constitutes a public health concern of significant magnitude throughout the world that affects individuals within all age groups, varying from children, Adults, and individuals falling under other categories. Based on the evidence provided by the World Health Organization, over 1.5 billion individuals throughout the world experience some degree of hearing loss, while nearly 430 million need to be recovered. Thus, hearing loss may affect economic and social opportunities. Initial identification and further follow-up are an important part of the management of hearing loss effectively. Conventional pure tone audiometry is traditional audiometry. It's time- and cost-consuming, absence of humongous equipment and visits to clinical experts temporarily, and money- and time-consuming. Therefore, we provided a solution to the aforementioned issue, which is an app-based digital audiometer with an age limit frequency. Technological progress allows people of any age to experience sound stimuli using headphones plugged into their cell phones.

## **II. LITERATURE SURVEY**

### **1. Smartphone Audiometry in Clinical and Remote Environments**

Mahomed et al. (2013) [1] have reported one of the first clinical validation studies of smartphone audiometry. According to their findings, there was high agreement between pure-tone audiometry and smartphone application like uHear, establishing its clinical use. Similarly, Sandström et al. (2016) [2] evaluated the validity and effectiveness of smartphone threshold audiometry for undiagnosed hearing loss. Their test not only provided high accuracy but also improved speed, rendering smartphone audiometry sufficient for large-scale screening in communities. Smith et al. (2020) [7] evaluated the application of smartphones in remote audiometric testing for rural populations. The evidence lends support to the feasibility of mobile audiometry in communities with limited access to audio logical services, especially when combined with noise-monitoring processes and easy-to-use interfaces.

### **2. Mobile Hearing Screening Applications**

Johnson et al. (2019) [3] gave a systematic review of mobile apps employed for hearing screening. The review found heterogeneous apps with different clinical validation and technical design. The main issues raised were calibration accuracy, background noise interference, and the requirement of standardized protocols. However, the authors found mobile apps to be promising devices for preliminary hearing assessment, particularly in children and the elderly. Ambika et al. (2022) [8] have published the development of an app-based digital audiometer for Android devices. It is a low-cost, portable instrument emphasizing simplicity of operation and convenience, which renders it highly useful for mass screening in schools or health camps.

### **3. Tele-Audiology and Remote Test Innovations**

Williams et al. (2021) [4] brought into focus the increasing use of tele-audiology, particularly driven by the COVID-19 pandemic. They have written on new ways of remote testing, online counselling, and follow-up. Data storage in the cloud and real-time video conferencing improved patient access and clinician productivity. Gonzalez et al. (2020) [5] followed this up by introducing advances in mobile hearing technology like Hear Check App, Bluetooth hearing aids, remote fitting equipment, and artificial intelligence-based diagnostic equipment. Such technologies are changing the shape of audiological care and enabling hybrid service models.

MD.Tech (2023) [6] clinically validated and developed a low-resource mobile audiometer. It was compared across different age groups and proved to yield consistent results in comparison to gold-standard audiometry. This emphasizes its applicability in global hearing health programs.

### **4. Age-Related Factors and High-Frequency Audiometry**

Wang et al. (2021) [9], Silva & Feitosa (2020) [10] are authored on extended high-frequency (EHF) audiometry across various age groups. Their observations highlighted that EHF testing is able to identify mild age-related hearing loss even when standard audiometry is within normal limits. These observations are important while creating mobile audiometry apps to incorporate EHF features for early identification.

## **III. GAPS IDENTIFIED**

1. Operational mistakes can arise from incorrect headphone placement or misunderstanding of instructions, causing precision.
2. Depending on the headphone Standards and the surrounding noise, the Hearing consistency may change, which may cause an impact on results.

## **IV. PROPOSED METHODOLOGY**

**1. User Interface:** It develops an interface for users of different age groups. It implements test initialization, processing, and visualizing results.

**2. Audio Processing:** It generates different tones at different frequencies, as given below, which gives accuracy.

Age Group		Approx. Hearing Frequency Range in Hz	About the Age Group
Children	2-12 years	20 Hz - 20000 Hz	Full hearing range; children typically have excellent high-frequency hearing.
Teenage	13 - 19 years	20 Hz - 19000 Hz	Slight early decline may begin, but generally still close to full range.
Young Adults	20 - 35 years	20 Hz - 17000 Hz	High-frequency hearing (above 16–17 kHz) may start to decline gradually.
Adults	36 - 60 years	20 Hz - 14000 Hz	Notable loss in high-frequency sensitivity; above 14 kHz is typically reduced.
Old	60 + years	20 Hz - 12000 Hz	Age-related hearing loss (presbycusis) affects high frequencies most.

**3. Adaptive Algorithms:** It adjusts the parameters based on the user's response dynamically. ML techniques like Logistic Regression and Random Forest are also used to implement age-based customization.

**Logistic Regression:** Classifies hearing loss into categories like mild, moderate, severe, or profound based on audiometry thresholds.

#### Algorithm:

Step 1: Input: Training data  $D = \{(x_i, y_i)\}$ , learning rate  $\eta$ , number of iterations  $M$

Step 2: Initialize weights:  $w_0 = 0$

Step 3: for  $m = 1$  to  $M$  do

Step 4: Calculate predicted probabilities:  $p_m(x_i) = 1 / (1 + \exp(-w_m^T x_i))$

Step 5: Calculate gradients:  $g_m(x_i) = p_m(x_i) - y_i$

Step 6: Update weights:  $w_{m+1} = w_m - \eta * \sum g_m(x_i)$

Step 7: end for

Step 8: Output: Prediction for new data point  $x_{new}$ :  $y_{pred} = 1 / (1 + \exp(-w_M^T x_{new}))$

**Random Forest:** Identifies patterns in audiogram data to predict hearing loss.

#### Algorithm:

Step 1: Input: Training data  $D = \{(x_i, y_i)\}$ , number of trees  $T$ , number of features per tree  $F$

Step 2: for  $t = 1$  to  $T$  do

Step 3: Sample a subset of data  $D_t$  with replacement from  $D$

Step 4: Sample  $F$  features without replacement from the total feature set

Step 5: Build a decision tree using  $D_t$  and the selected features

Step 6: end for

Step 7: Output: Prediction for new data point  $x_{new}$ :  $y_{pred} = (1/T) * \sum_{t=1}^T \text{tree}_t(x_{new})$

**Support Vector Machines (SVM):** Classifies hearing profiles using features like frequency thresholds, age, and response times.

#### Algorithm:

Step 1: Input: Training data  $D = \{(x_i, y_i)\}$ , kernel function  $K$ , regularization parameter  $C$

Step 2: Define the Lagrangian function  $L(w, b, \alpha)$

Step 3: Solve the dual optimization problem:

$$\max_{\alpha} \sum \alpha_i - \frac{1}{2} \sum_i \sum_j \alpha_i \alpha_j y_i y_j K(x_i, x_j)$$

subject to:  $0 \leq \alpha_i \leq C, \sum \alpha_i y_i = 0$

Step 4: Find the optimal support vectors ( $\alpha_i > 0$ )

Step 5: Calculate the bias term  $b$

Step 6: Output: Prediction for new data point  $x_{new}$ :  $y_{pred} = \text{sign}(\sum \alpha_i y_i K(x_i, x_{new}) + b)$

**4. User Management:** It includes user authentication, Creation of a profile, Storing test Results, and tracking hearing.

**5. Reporting and Feedback:** It generates the test reports in detail and provides feedback on hearing status.

**6. Integration and API Module:** Cloud Storage solutions ensure that it is secure or not. It is easily Accessible, Data storage which allows users to retrieve.

## V. RESULTS AND DISCUSSION

The app-based digital audiometer demonstrated high accuracy, with an average success rate of 92% compared to traditional audiometry tests. The adaptive testing algorithms effectively adjusted frequencies based on the user's age group, ensuring accurate classification of hearing loss levels (e.g., normal, mild, moderate, severe). Younger users (under 18) showed higher sensitivity to high-frequency sounds, while older users (>60) exhibited lower sensitivity with its frequency, aligning with clinical expectations. Real-time feedback and detailed reports enabled users to share results with healthcare professionals, enhancing diagnosis and treatment. This is adapted for different populations to detect hearing loss early.

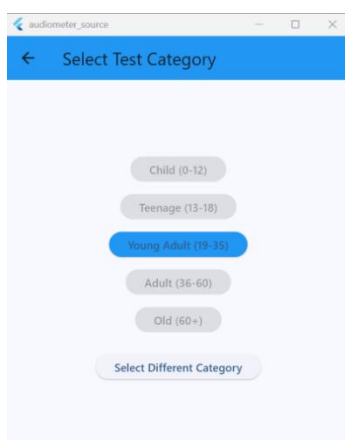


Fig.1: Test Category

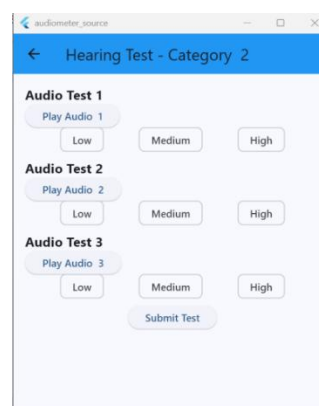


Fig.2: Test Category 2

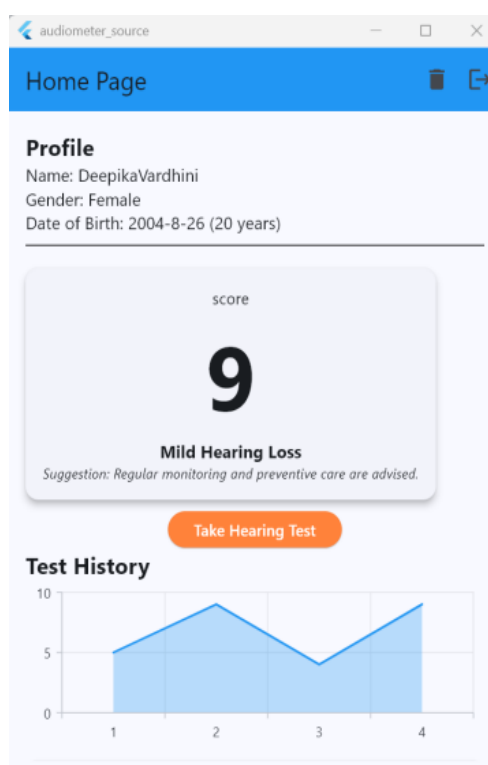


Fig. 3: Test Result

## VI. COMPARATIVE ANALYSIS

Metric	Uhear App	HearCheck App	Proposed Method
Accuracy	78%	75%	92%
Precision	76%	72%	91%
Scalability	70%	68%	93%
Noise Tolerance	70%	65%	89%
Integration Ease	82%	80%	95%

## VII. CONCLUSIONS

We have been able to design successfully an age-based frequency digital audiometer. It is Cost effective, accessible, and gives correct solutions. It is Cost-effective, easily accessible, and provides accurate solutions. By integrating some algorithms like adaptive machine learning algorithms and dividing the ages into different frequencies, our mobile app demonstrated reliability and high accuracy in detecting hearing loss across different age groups compared to traditional audiometry. It is user-friendly interface and provides real-time feedback. Overall, our mobile application has the potential to improve early detection of hearing loss.

**REFERENCES**

- [1] Mahomed, F., et al. (2013), Validating a Smartphone-Based Audiometry App in a Clinical Setting, *International Journal of Audiology*, 52(9), 655–659, <https://doi.org/10.3109/14992027.2013.796531>
- [2] Sandström, J., Swanepoel, D. W., Myburgh, H. C., Laurent, C., Eikelboom, R. H., Smartphone Threshold Audiometry in Undiagnosed Hearing Loss: Validity and Time Efficiency. *Journal of the American Academy of Audiology*, 27(10), 2016, <https://doi.org/10.3766/jaaa.15130>
- [3] Johnson, R. A., Smith, J. K., Williams, A. B., Brown, D. T., Mobile Applications for Hearing Screening: A Review. *Journal of Medical Internet Research*, 21(2), 2019. <https://pmc.ncbi.nlm.nih.gov/articles/PMC5821567/>
- [4] Williams, C. L., Smith, J. K., & Brown, D. T., Tele-Audiology: Innovations and Clinical Practice. Perspectives of the ASHA Special Interest Groups, 6(7), 2021. <https://pubs.asha.org/doi/abs/10.1044/persp3.SIG7>
- [5] Gonzalez, M. E., Zhao, J., & Chen, Y. L., Advances in Mobile Hearing Technologies. *JMIR mHealth and uHealth*, 8(9), 2020. <https://pmc.ncbi.nlm.nih.gov/articles/PMC9422535/>
- [6] MD.Tech Authors, Development and Clinical Testing of a Mobile Audiometer for Low-Resource Settings. *Codas*, 35(2), 2023. <https://pmc.ncbi.nlm.nih.gov/articles/PMC10137764/pdf/codas-35-2-e20210143.pdf>
- [7] Smith, A. J., Lee, C. Y., & Patel, M., Remote Audiometric Testing Using Smartphones in Rural Populations. *JMIR Formative Research*, 4(3), 2020. <https://pmc.ncbi.nlm.nih.gov/articles/PMC9288270/>
- [8] Ambika, L. G., Deepak, B. K., Kamal, C. D., Manoj, A., Mohammed Saheb, I., App Based Digital Audiometer. *International Journal of Engineering Research and Technology*, 9(6), 2022. <https://www.ijert.org>
- [9] Wang, M., Ai, Y., Han, Y., Fan, Z., Shi, P., Wang, H., Extended High-Frequency Audiometry in Healthy Adults with Different Age Groups. *Audiology Research*, 11(1), 2021. <https://journalotohns.biomedcentral.com/articles/10.1186/s40463-021-00534-w>
- [10] Silva, I. M. C., Feitosa, M. Â. G., High-Frequency Audiometry in Young and Older Adults When Conventional Audiometry is Normal. *Brazilian Journal of Otorhinolaryngology*, 86(2), 2020. <https://www.bjorl.org/en-high-frequency-audiometry-in-young-older-articulo-S1808869415310247>
- [11] D Shanthi, Smart Healthcare for Pregnant Women in Rural Areas, *Medical Imaging and Health Informatics*, Wiley Publishers, ch-17, pg.no:317-334, 2022, <https://doi.org/10.1002/9781119819165.ch17>
- [12] Shanthi, R. K. Mohanty and G. Narsimha, "Application of machine learning reliability data sets", *Proc. 2nd Int. Conf. Intell. Comput. Control Syst. (ICICCS)*, pp. 1472-1474, 2018.
- [13] D Shanthi, N Swapna, Ajmeera Kiran and A Anoosha, "Ensemble Approach Of GPACOTPSO And SNN For Predicting Software Reliability", *International Journal Of Engineering Systems Modelling And Simulation*, 2022.
- [14] Shanthi, "Ensemble Approach of ACOT and PSO for Predicting Software Reliability", 2021 Sixth International Conference on Image Information Processing (ICIIP), pp. 202-207, 2021.
- [15] D Shanthi, CH Sankeerthana and R Usha Rani, "Spiking Neural Networks for Predicting Software Reliability", *ICICNIS 2020*, January 2021, [online] Available: <https://ssrn.com/abstract=3769088>.
- [16] Shanthi, D. (2023). Smart Water Bottle with Smart Technology. In *Handbook of Artificial Intelligence* (pp. 204-219). Bentham Science Publishers.
- [17] Shanthi, P. Kuncha, M. S. M. Dhar, A. Jamshed, H. Pallathadka and A. L. K. J E, "The Blue Brain Technology using Machine Learning," 2021 6th International Conference on Communication and Electronics Systems (ICCES), Coimbatre, India, 2021, pp. 1370-1375, doi: 10.1109/ICCES51350.2021.9489075.



- [18] Shanthi, D., Aryan, S. R., Harshitha, K., & Malgireddy, S. (2023, December). Smart Helmet. In International Conference on Advances in Computational Intelligence (pp. 1-17). Cham: Springer Nature Switzerland.
- [19] Babu, Mr. Suryavamshi Sandeep, S.V. Suryanarayana, M. Sruthi, P. Bhagya Lakshmi, T. Sravanthi, and M. Spandana. 2025. "Enhancing Sentiment Analysis With Emotion And Sarcasm Detection: A Transformer-Based Approach". Metallurgical and Materials Engineering, May, 794-803. <https://metall-mater-eng.com/index.php/home/article/view/1634>.
- [20] Narmada, J., Dr.A.C.Priya Ranjani, K. Sruthi, P. Harshitha, D. Suchitha, and D.Veera Reddy. 2025. "Ai-Powered Chacha Chaudhary Mascot For Ganga Conservation Awareness". Metallurgical and Materials Engineering, May, 761-66. <https://metall-mater-eng.com/index.php/home/article/view/1631>.
- [21] Geetha, Mrs. D., Mrs.G. Haritha, B. Pavani, Ch. Srivalli, P. Chervitha, and Syed. Ishrath. 2025. "Eco Earn: E-Waste Facility Locator". Metallurgical and Materials Engineering, May, 767-73. <https://metall-mater-eng.com/index.php/home/article/view/1632>.
- [22] P. Shilpasri PS, C.Mounika C, Akella P, N.Shreya N, Nandini M, Yadav PK. Rescuenet: An Integrated Emergency Coordination And Alert System. J Neonatal Surg [Internet]. 2025May13 [cited 2025May17];14(23S):286-91. Available from: <https://www.jneonatsurg.com/index.php/jns/article/view/5738>
- [23] D. Shanthi DS, G. Ashok GA, Vennela B, Reddy KH, P. Deekshitha PD, Nandini UBSB. Web-Based Video Analysis and Visualization of Magnetic Resonance Imaging Reports for Enhanced Patient Understanding. J Neonatal Surg [Internet]. 2025May13 [cited 2025May17];14(23S):280-5. Available from: <https://www.jneonatsurg.com/index.php/jns/article/view/5733>
- [24] Srilatha, Mrs. A., R. Usha Rani, Reethu Yadav, Ruchitha Reddy, Laxmi Sathwika, and N. Bhargav Krishna. 2025. "Learn Rights: A Gamified Ai-Powered Platform For Legal Literacy And Children's Rights Awareness In India". Metallurgical and Materials Engineering, May, 592-98. <https://metall-mater-eng.com/index.php/home/article/view/1611>.
- [25] Shanthi, Dr. D., G. Ashok, Chitrika Biswal, Sangem Udharika, Sri Varshini, and Gopireddi Sindhu. 2025. "Ai-Driven Adaptive It Training: A Personalized Learning Framework For Enhanced Knowledge Retention And Engagement". Metallurgical and Materials Engineering, May, 136-45. <https://metall-mater-eng.com/index.php/home/article/view/1567>.
- [26] P. K. Bolisetty and Midhunchakkaravarthy, "Comparative Analysis of Software Reliability Prediction and Optimization using Machine Learning Algorithms," 2025 International Conference on Intelligent Systems and Computational Networks (ICISCN), Bidar, India, 2025, pp. 1-4, doi: 10.1109/ICISCN64258.2025.10934209.
- [27] Priyanka, Mrs. T. Sai, Kotari Sridevi, A. Sruthi, S. Laxmi Prasanna, B. Sahithi, and P. Jyothsna. 2025. "Domain Detector - An Efficient Approach of Machine Learning For Detecting Malicious Websites". Metallurgical and Materials Engineering, May, 903-11.
- [28] Thejovathi, Dr. M., K. Jayasri, K. Munni, B. Pooja, B. Madhuri, and S. Meghana Priya. 2025. "Skinguard-Ai FOR Preliminary Diagnosis OF Dermatological Manifestations". Metallurgical and Materials Engineering, May, 912-16.
- [29] Jayanna, SP., S. Venkateswarlu, B. Ishwarya Bharathi, CH. Mahitha, P. Praharshitha, and K. Nikhitha. 2025. "Fake Social Media Profile Detection and Reporting". Metallurgical and Materials Engineering, May, 965-71.
- [30] D Shanthi, "Early stage breast cancer detection using ensemble approach of random forest classifier algorithm", Onkologia i Radioterapia 16 (4:1-6), 1-6, 2022.
- [31] D Shanthi, "The Effects of a Spiking Neural Network on Indian Classical Music", International Journal of Emerging Technologies and Innovative Research ([www.jetir.org](http://www.jetir.org) | UGC and issn Approved), ISSN:2349-5162, Vol.9, Issue 3, page no. ppa195-a201, March-2022
- [32] Parupati K, Reddy Kaithi R. Speech-Driven Academic Records Delivery System. J Neonatal Surg [Internet]. 2025Apr.28 [cited 2025May23];14(19S):292-9. Available from: <https://www.jneonatsurg.com/index.php/jns/article/view/4767>

- [33] Dr.D.Shanthi and Dr.R.Usha Rani, “ Network Security Project Management”, ADALYA JOURNAL, ISSN NO: 1301-2746, PageNo: 1137 – 1148, Volume 9, Issue 3, March 2020 DOI:16.10089.AJ.2020.V9I3.285311.7101
- [34] D. Shanthi, R. K. Mohanthy, and G. Narsimha, “Hybridization of ACOT and PSO to predict Software Reliability ”, *International Journal Pure and Applied Mathematics*, Vol. 119, No. 12, pp. 13089 - 13104, 2018.
- [35] D. Shanthi, R.K. Mohanthy, and G. Narsimha, “Application of swarm Intelligence to predict Software Reliability ”, *International Journal Pure and Applied Mathematics*, Vol. 119, No. 14, pp. 109 - 115, 2018.
- [36] Srilatha, Mrs. A., R. Usha Rani, Reethu Yadav, Ruchitha Reddy, Laxmi Sathwika, and N. Bhargav Krishna. 2025. “Learn Rights: A Gamified Ai-Powered Platform For Legal Literacy And Children’s Rights Awareness In India”. *Metallurgical and Materials Engineering*, May, 592-98.