

The Sounds of Home: A Speech-Removed Residential Audio Dataset for Sound Event Detection





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Introduction 1

- Smart home systems offer opportunities to enhance older adults' wellbeing through audio machine learning, but require substantial labeled data capturing Activities of Daily Living (ADL).
- We present **The Sounds of Home**, a comprehensive residential audio dataset for sound event detection research in smart home applications, in collaboration with LiCalab as part of the Vitalise project https://vitalise-project.eu/
- Our dataset of 1342 hours length was constructed by deploying audio recording systems in 8 participants' homes (ages 55-80) for 7 days, with a novel automated speech removal pipeline to address privacy concerns.
- The resulting privacy-compliant dataset accurately captures residential soundscapes, enabling development and benchmarking of sound event detection models for in-home applications.

OBJECTIVES

- Create a large-scale, privacy-compliant audio dataset capturing real-world domestic soundscapes of older adults for sound event detection research.
- Develop and implement an automated speech removal pipeline to ensure participant privacy while preserving nonspeech audio events.

DATASET COLLECTION METHODOLOGY







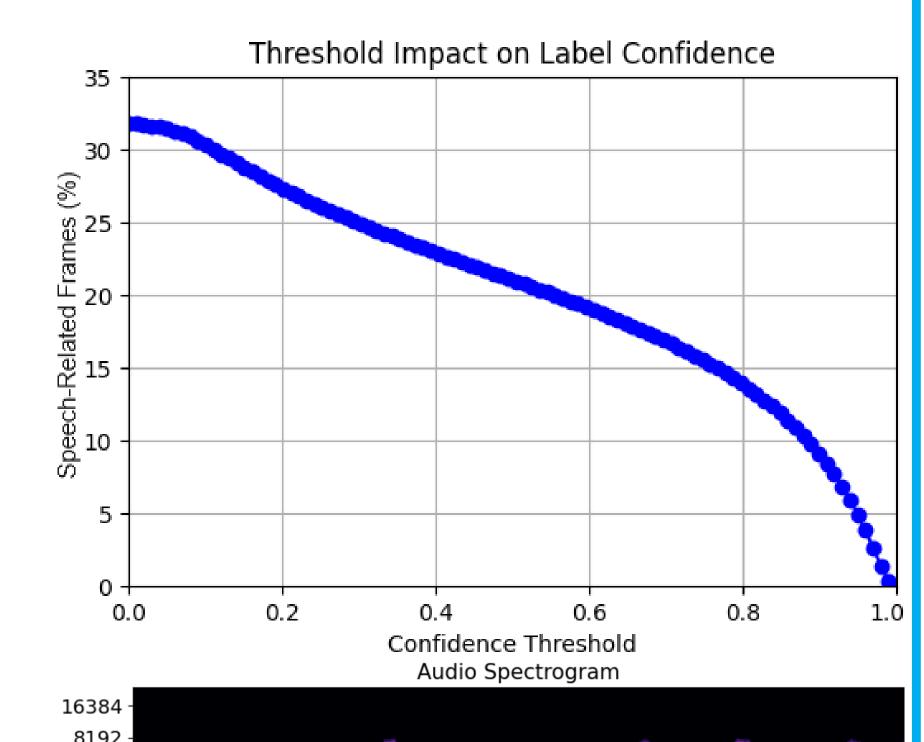


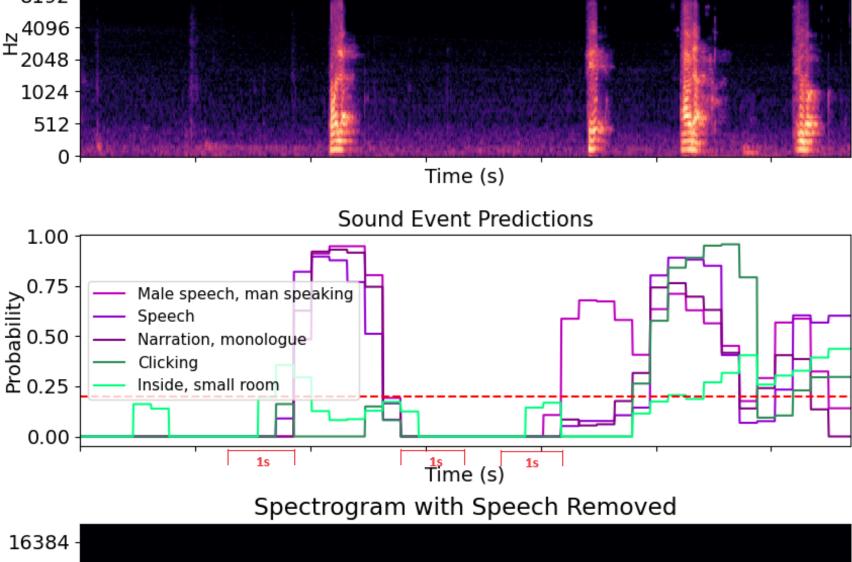
- Deployed two Audio Recording Systems (ARS) in each of 8 participants' homes for 7 days, using AudioMoth devices enclosed in IPX7 Waterproof Cases to ensure durability and high-quality audio capture.
- Configured devices to record WAV audio at 48 kHz sample rate in 1-hour segments from 8:00 AM to 9:00 PM daily, capturing participants' most active periods while minimizing interference with daily routines.
- Placed devices strategically in living rooms and kitchens at 1-2 meters height, away from windows and obvious noise sources, with vibration pads to reduce ground-borne vibration impact.
- Documented acoustic characteristics through detailed floor plans, construction material information, and placement photographs to enable future replication of recording environments.

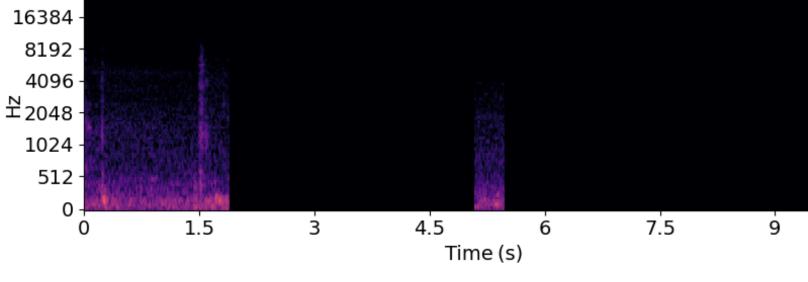
DATASET ANONYMISATION METHODOLOGY

Raw Feature PANNs Threshold Speech Analysis Speech Audio

- 1. AI Models for Inference
 - Employed Pretrained Audio Neural Networks (PANNs)
 - Used two specific models: Cnn14_DecisionLevelAtt and Cnn14_DecisionLevelMax
 - Models pre-trained on AudioSet with 48% speech-labeled data
- 2. Feature Extraction and Analysis
 - Extracted log-mel spectrogram features (frame size: 1024 samples, hop size: 512 samples)
 - Analyzed 32 frames at a time for inference
 - Saved the 7 detection results with most confidence in JSON format, including frame indices and probabilities
- 3. Threshold Analysis and Speech Removal
 - Focused on distribution of predefined vocal-related labels
 - Set confidence threshold at 0.2 based on statistical analysis and manual verification
 - Removed segments containing speech-related labels above threshold
 - Extended removal intervals by one second before and after detected speech
 - Speech-related labels considered: Speech, Singing, Male singing, Female singing, Child singing, Male speech, man speaking, Female speech, woman speaking, Conversation, Narration, monologue, Music
 - Replacement of identified speech segments with low-amplitude noise (1e-10).







ACCESS TO DATASET AND ANONYMIZATION SOFTWARE

In the following link you will find: 1. The "The Sounds of Home" dataset, 2. The software used to remove the sections containing spoken voice and 3. A Jupiter Notebook uploaded to Google Colab where you can analyze your own audio files using our software. Just upload the file and you're done!



This work was supported by Engineering and Physical Sciences Research Council (EPSRC) Grant EP/T019751/1 AI for Sound. AI4S project link: https://ai4s.surrey.ac.uk/index.

https://github.com/gbibbo/voice_anonymization

