Visual Sound Assistant

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Goal

❑ **Goals ---- Helping people with hearing loss**

- Aiming to help people suffering from hearing loss, we developed an Android application that can collect sound information.
- Our App extracts the information hidden in the background sounds and convert it into text. If something dangerous is happening, our App will notify users on time.

Motivations and Objectives

❑ **Motivations**

- Apply **deep learning** and **mobile technology** in hearing assistance area.
- Various sound recognition.
- Help people with hearing issues to have the environmental sound sensitivity.

❑ **Technical Objectives**

- An **Android application** to identify sound classification.
- Collect **sound information around** like barking, shooting, alarming, etc...
- Send **real-time notifications** to users.

Research Challenges

❑ Sound signal processing algorithms.

❑ Neural network training with limited datasets.

❑ Noises can interfere the accuracy.

❑ Duration of the audio sample also can affect the accuracy.

❑ Manually customize server. Updating and synchronizing the neural network model between clients and server.

❑ User experience issues on Android.

Acknowledgement

We would like to thank Prof. Yingying Chen for her kind support and guidance through the whole program.

Methodology

Sound Signal Processing

❑ MFCC(Mel-Frequency Cepstral Coefficients) based on LibROSA[1] extracts 40 features per audio.

Neural Network Structure

❑ A fully connected neural network built on TensorFlow.

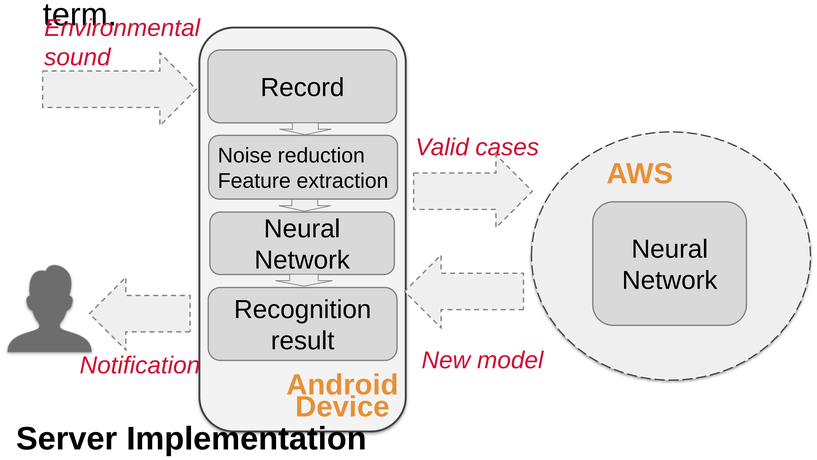
❑ Classify up to 10 categories using UrbanSound8k[2] dataset.

Android Development

❑ Multi-thread audio recorder and background service ensure zero omission.

❑ Adjustable notification priority to avoid annoyance.

❑ Interactions with the server can effectively collect cases and achieve more robust model in the long term.



Server Implementation

❑ Serve a TensorFlow Model online.

❑ The server is based on Amazon Web Service(AWS).

❑ The server receives valid cases from App, continuously trains the neural network and send the new model to App.

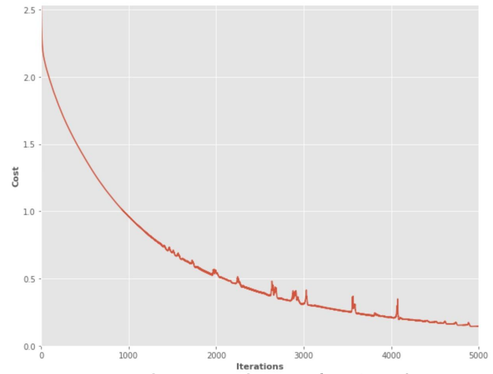
Results

❑ Extract **40** MFCC features.

❑ **10** Classification categories.

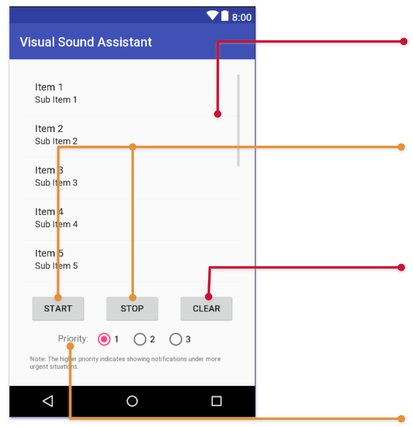
❑ Accuracy over **96%** on UrbanSound8k.

❑ Real-time response **within one second.**



F-Score: 0.852

f-score of neural network



Classification results are shown in the textbox

"Start" & "Stop" Buttons to start/stop sound recognition

"Clear" Button to clear the textbox above

User can choose different notification priorities

Future Development

❑ Hot-update features.

❑ More layers to enhance the recognition accuracy.

❑ Move software to a smaller Linux device and iOS.

References

[1] LibROSA: https://github.com/librosa/librosa
[2] UrbanSound8k: https://serv.cusp.nyu.edu/projects/urbansounddataset/urbansound8k.html
[3] Piczak, Karol J. "Environmental sound classification with convolutional neural networks." Machine Learning for Signal Processing (MLSP), 2015 IEEE 25th International Workshop on. IEEE, 2015.