Speaker Identification in a Noisy Environment Raw Waveforms vs MFCC





Introduction

- We built a Speaker Identification network.
 Input is a voice sample, output is who's talking.
- Our motivation came from Alzheimer's patients
 who forget or can't identify who is speaking in
 group calls, e.g. family calls to Grandma from
 one speaker phone.
- The solution extends to voice conference calls, as seen on right.
- Many existing solutions perform Speaker Identification only on clean audio. We use noisy audio to better simulate real-world conditions.
- Our network is trained on 20 speakers, as that's roughly the max size of a family or work team.





Datase

20 audiobooks from LibriVox

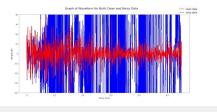
CHARLES DARWIN



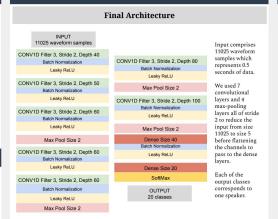
- LibriVox and Internet Archive host royalty-free audiobooks read by volunteers.
- Mono-channel, 22050Hz mp3s; one file per chapter.
- We split into 0.5 second training examples.. 0.5 seconds seemed to be the shortest time a human needs to identify the speaker.
- Complication: Some audio books have multiple speakers. Some speakers read multiple audio books.

Augmenting with Noise

- Audiobooks are clean recordings. We want background noise.
- 3 Background Noises: 1) Crowd Talking, 2) Laptop Keyboard, 3) Plastic Crumple
- Overlay ENTIRE audiobook with noise in 20-second chunks. Normalize volume to match.

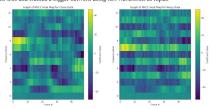


Algorithms and Models



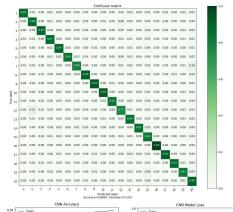
Experiments

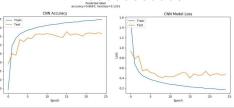
- Model trained on clean data achieves 94% test accuracy on clean data.
- Clean model achieves a paltry 14% accuracy when making predictions on noisy data.
- Significant effort training models with MFCCs, which are generated from the raw
 waveform via substantial signal preprocessing. Traditionally, most Speech Recognition and
 Speaker Identification use MFCCs as input.
- The MFCC models achieved 88% accuracy on clean data, but only 63% on noisy data.
- TA advised that the preprocessing could be discarding valuable signal, so we abandoned MFCCs and trained a bigger network using raw waveforms as input.



Results & Analysis







References

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