Activity Recognition Using Both Computer Vision and Audio Processing



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• How do you recognize an acquaintance?



- How do you recognize an acquaintance?
 - by voice and appearance



- How do you recognize an acquaintance?
 - by voice and appearance
- What about machines?



- How do you recognize an acquaintance?
 - by voice and appearance
- What about machines?
 - with our findings in the machine learning field and labeled data, will machines have the ability to recognize the speakers?



Applications

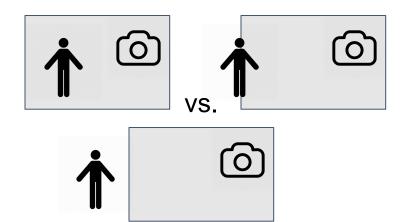
- Personalized Smart Assistants (Alexa, Google Home, Siri...)
- Smart Baby Monitors
- Emotion Detection
- Image Captioning



Task

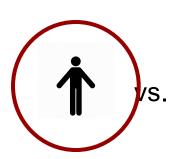
Given audio and visual data of a person talking, can we identify who it is?



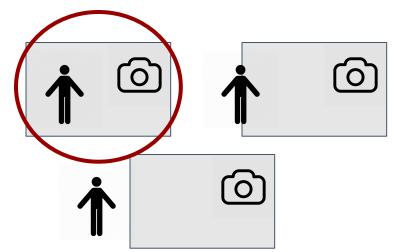


Task

Given audio and visual data of a person talking, can we identify who it is?







Data Collection:

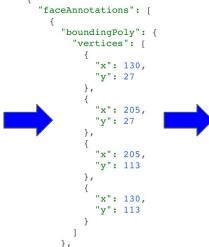
- Background Variation:
 - 5 different backgrounds
- Sentence Variation:
 - 10 different sentences
- Clothing Variation:
 - 2 different clothing (jacket/no jacket)
- Total Data Collected (10 people):
 - Video Samples: 188
 - Audio Samples: 188
 - Image Samples: 16,668



Visual Recognition



Training data: sampled frames from videos labeled with correct person in video



Output from Google Cloud Vision API



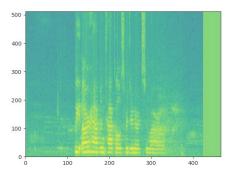
Standard size JPEG image for neural network



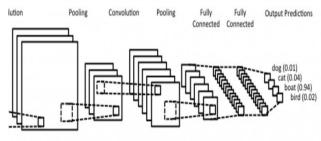
Determines most likely person in a given picture by comparing it to the training data

Audio Recognition

- Collecting data
- Converting mp3 to wav file
- Loaded wav file data to Google Colab
- Converting wav file to spectrograms
- Shuffle, batch and normalize the spectrograms
- Build Convolutional Neural Network
- Parameter tuning Learning rate



Spectrogram of Bella's Voice



Convoluted Neural Network





Preliminary Results

- Computer Vision:
 - Classification Accuracy:
 - Frame Classification Model: 45%
 - Classifies every frame in the video
 - Video Classification Model: 60%
 - Groups the frames and weight the videos based on the classification of the Group
- Audio Processing:
 - Training Model: Hasn't been applied to the test data (set)
 - Minimum data loss: .286 from Audio Training Model

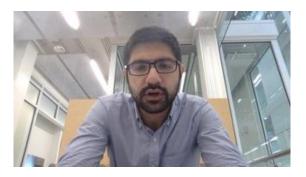




Preliminary Results

Confusion(Error) Matrix:

	Alyssa	Bardia	Bella	Fiona	Kara	Regina	Sancharz	Sophie	Trang	Violet
Alyssa	2									
Bardia		2								
Bella				2						
Fiona					2					
Kara					2					
Regina		2								
Sancharz		1					1			
Sophie								2		
Trang				1					1	
Violet				·						2



Bardia (Graduate Student)



Me

Future work

- 1. Improvement on Image Classification
 - a. Aggregate information from a sequence of frames, such as motion
 - b. Method for improving classification accuracy, i.e. cropping faces
 - c. Identify multiple people in images
- 2. Improvement on Speaker Identification
 - a. Finish Training
 - b. Audio to identify speakers from multi-speakers
- 3. Combine audio and video information to get better results
- 4. Be able to identify strangers in video

