

Capstone Project – 5

Verbal Communication Quality Monitoring & Feedback System [Speech Emotion Recognition]

Team Members

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Introduction



Q.1) What is Speech Emotion Recognition?

Speech Emotion Recognition (SER) can be defined as extraction of the emotional state of the speaker from his or her speech signal

Q.2) Why it is important?

Nowadays, more and more intelligent systems are using emotion recognition models to improve their interaction with humans. This is important, as the systems can adapt their responses and behavioral patterns according to the emotions of the humans and make the interaction more natural.

Q.3) What are the scopes of Speech Emotion Recognition? Security, medicine, entertainment, education etc.



Problem Statement

In this project we have to detect a person's emotions just by their voice which will let us manage many AI related applications. Detecting emotions is one of the most important marketing strategy in today's world. You could personalize different things for an individual specifically to suit their interest.

AIM:- We will solve the above-mentioned problem by applying deep learning algorithms to audio/speech data. The solution will be to identify emotions in speech.

Data Summary

disgust, 08 = surprised).



We have built a deep learning model which detects emotions of speech. For this purpose we've used Ryerson Audio-Visual Database of Emotional Speech and Song (RAVDESS)-2018 dataset. The description of dataset is as follows:

- RAVDESS contains 1440 files: 60 trials per actor x 24 actors = 1440. The RAVDESS contains 24 professional actors (12 female, 12 male)
- Speech emotions includes 8 emotions such as neutral, calm, happy, sad, angry, fearful, surprise, and disgust expressions.
- Filename identifiers: Modality (01 = full-AV, 02 = video-only, 03 = audio-only).
 Vocal channel (01 = speech, 02 = song).
 Emotion (01 = neutral, 02 = calm, 03 = happy, 04 = sad, 05 = angry, 06=fearful, 07 =

Ex filename = 03-01-06-01-02-01-12.wav = Audio-only (03), Speech (01), Fearful (06).



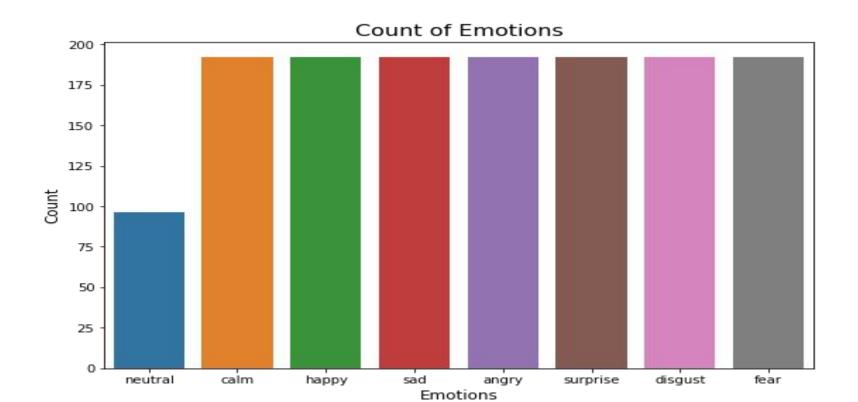
Data Preprocessing

- Extracting features using LIBROSA (librosa is a python package for music and audio analysis).
- Converting audio files into waveforms and spectrograms.
- Creating dataframe for extracted features for audio files.
- Data augmentation.

Exploratory Data Analysis



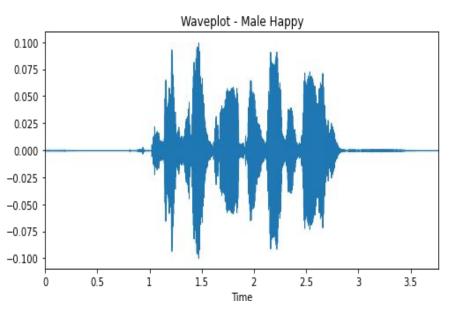
How many emotions are there for each class in audio dataset?

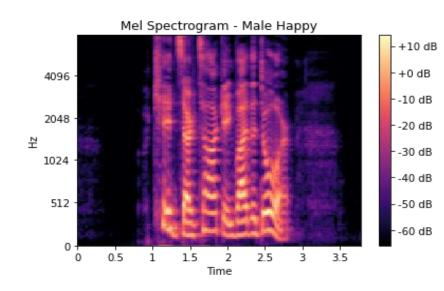


EDA(continued)



- A waveform is (mathematics) the shape of a wave function represented by a graph showing some dependent variable as function of an independent variable.
- A spectrogram is a visual representation of the spectrum of a sound changing through time.

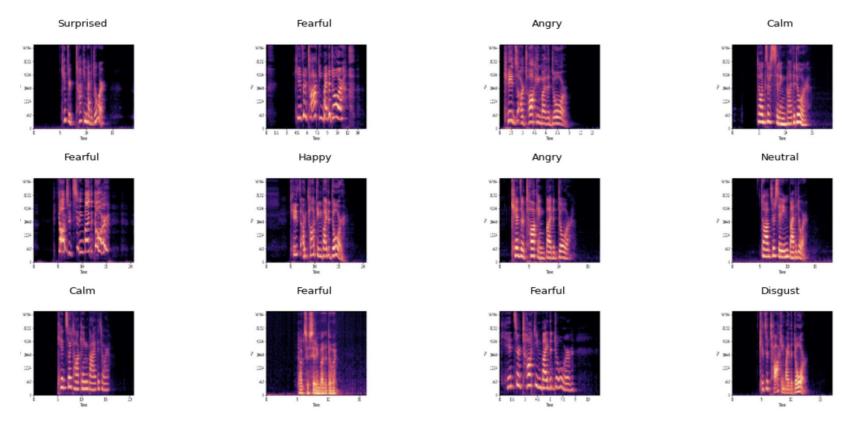




EDA(continued)



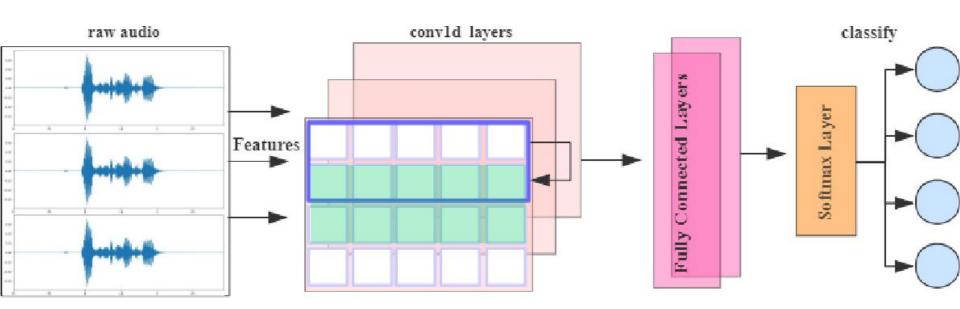
How different emotions looks in spectrogram?





Model implementation

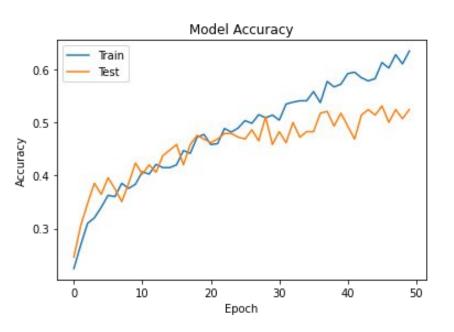
Base model CONVID:



Model implementation (Conv1D contd..)



Base model Conv1D performance:

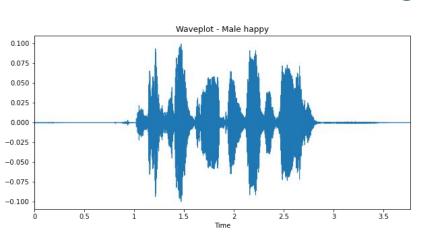


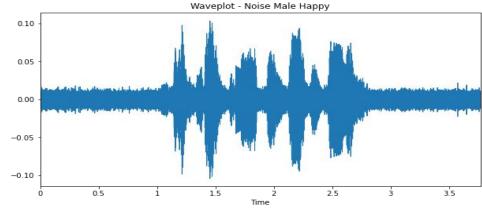
precision	recall	f1-score	support	
angry	0.50	0.51	0.51	39
calm disgust	0.61 0.71	0.74 0.68	0.67 0.69	38 37
fear	0.73	0.21	0.33	38
happy neutral	0.38 0.45	0.51 0.54	0.43	35 24
sad	0.40	0.47	0.43	38
surprise	0.62	0.54	0.58	39
accuracy			0.52	288
macro avg weighted avg	0.55 0.56	0.53 0.52	0.52 0.52	288 288

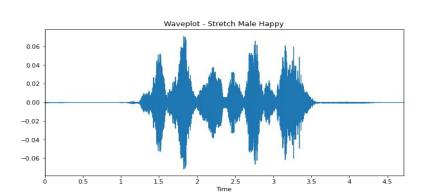
Model implementation (contd..)

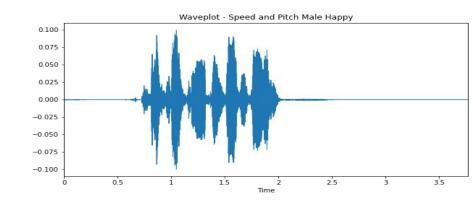


How audio files looks after **data augmentation**?





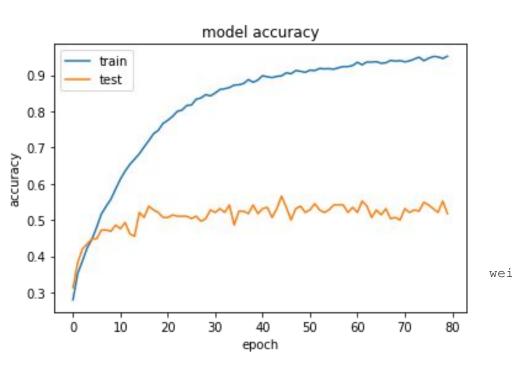




Model implementation(Conv1D contd...)



Implementing our base model Conv1D on augmented data:-

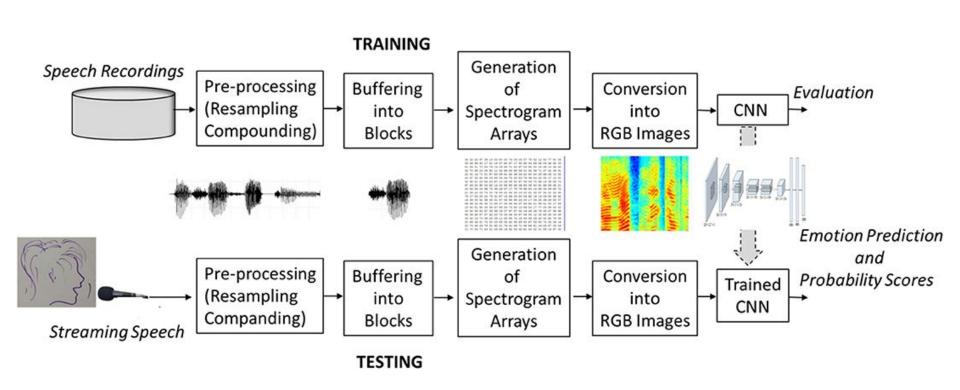


	precision	recall	f1-score	support
angry	0.58	0.56	0.57	39
calm	0.48	0.68	0.57	38
disgust	0.51	0.73	0.60	37
fear	0.59	0.42	0.49	38
happy	0.50	0.49	0.49	35
neutral	0.42	0.58	0.49	24
sad	0.47	0.21	0.29	38
surprise	0.59	0.49	0.54	39
accuracy			0.52	288
macro avg	0.52	0.52	0.50	288
ighted avg	0.52	0.52	0.51	288

Model Implementation (VGG-16 contd..)



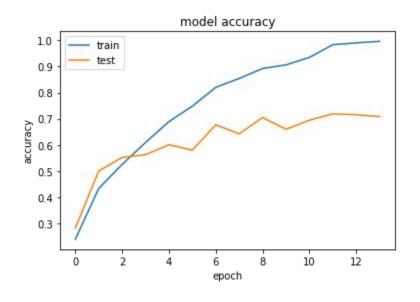
We then used transfer learning concept with VGG-16 Model



Model implementation (VGG-16 contd..)



VGG-16 performance:

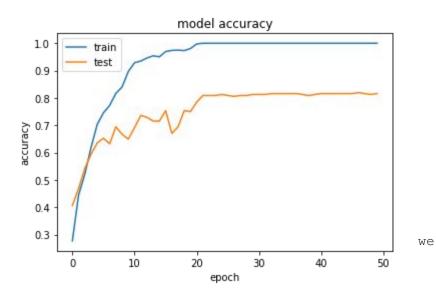


precision	recall	f1-score	support	
angry	0.87	0.67	0.75	39
calm	0.81	0.79	0.80	38
disgust	0.72	0.76	0.74	37
fearful	0.71	0.71	0.71	38
happy	0.54	0.63	0.58	35
neutral	0.65	0.71	0.68	24
sad	0.57	0.55	0.56	38
surprised	0.82	0.85	0.84	39
accuracy			0.71	288
macro avg	0.71	0.71	0.71	288
weighted avg	0.72	0.71	0.71	288

Model implementation (VGG-19 contd..)



VGG-19 (Fine Tuning) performance:

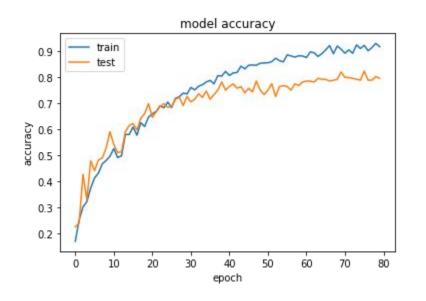


	precision	recall	f1-score	support	
aı	ngry	0.93	0.72	0.81	39
(calm	0.87	0.87	0.87	38
dis	gust	0.82	0.89	0.86	37
fea	rful	0.82	0.71	0.76	38
ha	арру	0.71	0.86	0.78	35
neu:	tral	0.81	0.88	0.84	24
	sad	0.72	0.76	0.74	38
surpr	ised	0.87	0.87	0.87	39
accu	racy			0.82	288
macro	avg	0.82	0.82	0.82	288
eighted	avg	0.82	0.82	0.82	288

Model implementation (VGG-19 contd..)



VGG-19 (Fine Tuning + Data Augmentation) performance:



р	recision	recall	f1-score	support	
angr	ΣΥ	0.78	0.82	0.80	39
cal	.m	0.76	0.89	0.82	38
disgus	st	0.77	0.92	0.84	37
fearfu	ıl	0.90	0.71	0.79	38
happ	ЭΥ	0.79	0.66	0.72	35
neutra	1	0.86	0.75	0.80	24
sa	ıd	0.77	0.61	0.68	38
surprise	ed	0.79	0.97	0.87	39
accurac	ΣÀ			0.80	288
macro av	rg	0.80	0.79	0.79	288
weighted av	rg	0.80	0.80	0.79	288

Model implementation (contd..)



We saved our implemented models with best values using checkpoints

Models (CNN)	Train accuracy	Test accuracy
Conv 1D	0.95	0.51
VGG-16 Fine tuning	0.98	0.71
VGG-19 Fine tuning	1.00	0.81
VGG-19 (Fine tuning + Augmentation)	0.91	0.82
Inception	0.97	0.73

Deployment



Tools used in deployment of deep learning project:-

Streamlit- Streamlit is an open-source python library that is useful to create and share data web apps.

Streamlit

GCP - Google Cloud Platform, commonly referred to as GCP, is a cloud computing service created by Google for building, testing, deploying, and managing applications and services through Google-managed data centers.

Google Cloud Platform

Deployment(contd..)



Implementation video of our project with a web app.

URL- https://ser-app-324611.ue.r.appspot.com/



Challenges

- As this project is not a stand-alone data science project, it required little to medium domain knowledge. Understanding Mfccs, Mel scale is very important for feature selection.
- Limitations include not using feature selection to reduce the dimensionality of our augmented CNN which may have improved learning performance.

 During deployment in Heroku and Azure, there were problems as some additional libraries needed to installed. After installing these libraries, some of previous requirements had to be downgraded in order to be compatible with the installed ones, that's why we deployed on GCP.



Conclusion

- VGG19 (fine tuning + augmentation) gave the best accuracy score of 82% and solved the problems like overfitting to some extent.
- It's quite difficult to get the accuracy of more than 90% due to lack of data.
- To solve problems like over-fitting that we had seen in almost every model, we need more real time data.
- Noise Adding, Pitching and Shifting for the imbalanced data was helping in achieving a better result.
- Computational cost was very high resulting in several runtime crashes but we're able to get our best model for deployment.



Thank You

