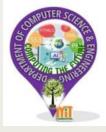


MAHARAJA INSTITUTE OF TECHNOLOGY THANDAVAPURA DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING



ANALYSING SPEECH DATA TO DETERMINE EMOTIONS

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Abstract

- The human voice is very versatile and carries a multitude of emotions. Emotion in speech carries extra insight about human actions.
- Through further analysis, we can better understand the motives of people, humans are easily able to determine the emotion of a speaker, but the field of emotion recognition through machine learning.
- The study of emotion in speech by detecting one emotion.

Introduction

The main objectives are following:

- The ability to understand vocal sounds.
- The human voice can be characterized by several attributes such as pitch, loudness, and vocal tone.
- humans express their emotions by varying different vocal attributes during speech generation.
- The human voice frequency is specifically a part of the human sound production mechanism.
- In which the vocal cords or folds are the primary source of generated sounds.

Literature Survey

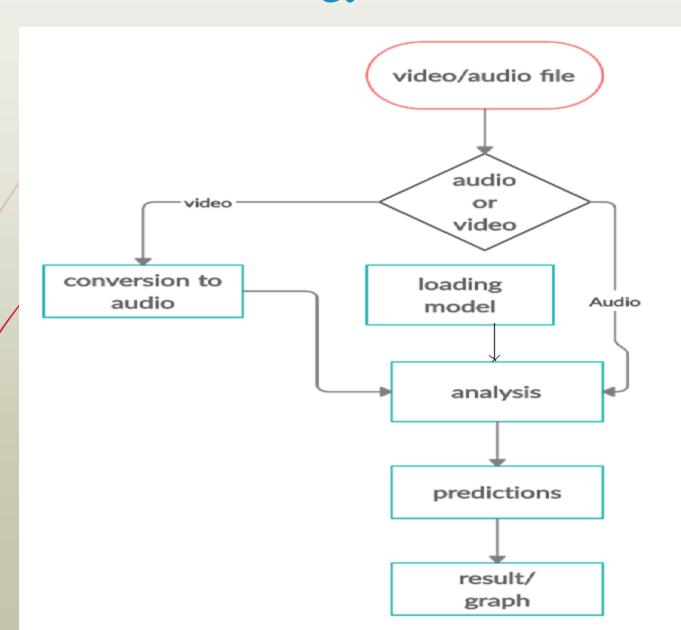
Year	Author	Title	Method	Dataset	Results	Drawbacks
2018	Maisy Wieman, Andy Sun Mithesh chouhan	Analyzing Vocal Patterns to Determine Emotion Speech- Emotion- Analyzer,	We investigate the best algorithms to select features that are relevant to predicting emotion. Neural networks	Stanford University Linguistics Department has an Emotional Prosody and Speech Corpus. Berlin voice dataset	Predicted with relative accuracy. (70%). Detect emotions with more than 70% accuracy.	Low accuracy, collection of data set is a tidiest task. Training takes a lot of time. Building the model was a challenging task as it involved lot of trail and error methods, tuning etc.
2007	Kamran Soltani Raja Noor Ainon	Speech emotion detection based on neural	Back propagation neural network as classifier.	Berlin Database	60% accuracy	Only 2 emotions were able to detect.

Continue...

2005	Chul Min	Toward	Linear	Real time	classifier	Very time
	Lee,	Detecting	discriminant	dataset	yielded	consuming
	Student	Emotions in	classifiers		promising	
	Member,	Spoken	(LDC) with		results with	
	IEEE, and	Dialogs	Gaussian		70%	
	Shrikanth S.		class-		classification	
	Narayanan,		conditional		accuracy	
	Senior		probability			
	Member,		and k-nearest			
	IEEE		neighborhood			
			classifiers (k-			
			NN).			
2011	Jangsik Cho	Detecting	Bayesian	Stand ford	Classifier	Taking
	,Shohei	emotion from	pairwise	voice	with 60%	probability
	Kato	voice	classifiers	dataset	prediction.	into
						consideration
						the ratio of
						correctness
						mainly
						depends on
						datasets.

22-04-2020

Methodology



Existing System

- The system that is currently present is google speech which also lacks in understanding the words when pronounced.
- It is time consuming.

Proposed System

- The system that is being proposed is to identify the modulation of the word in the speech data by understanding the tone in the voice that is being used.
- This helps in understanding whether the person is angry, happy, sad, neutral, calm, fearful, disgust, surprise etc.

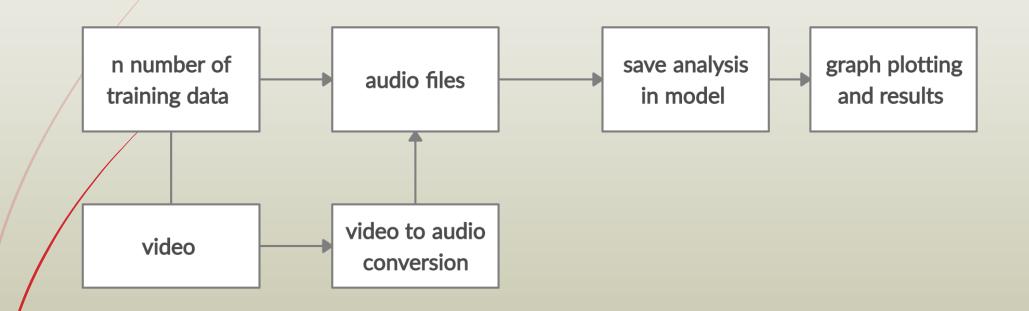
Functional Requirements

- Video input
- Audio input
- Predicted to emotion output

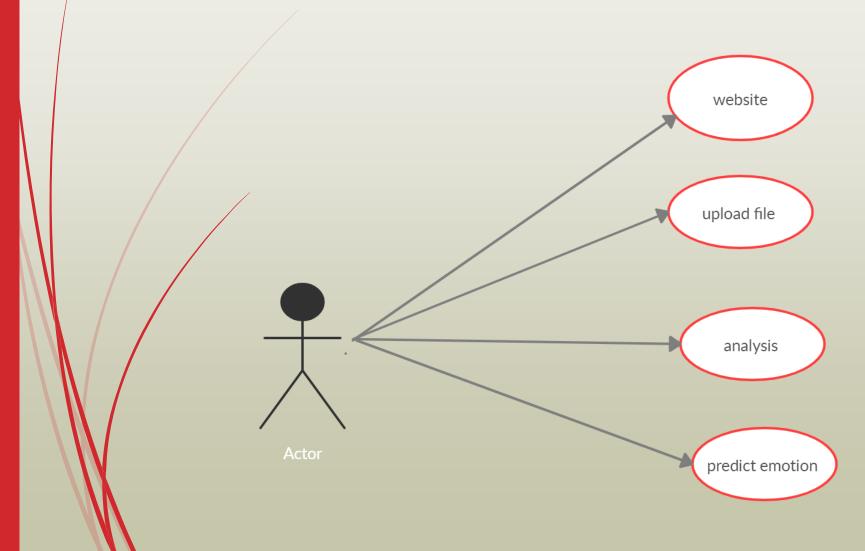
Non Functional Requirements

- Accessibility
- Documentation
- Maintainability
- Portability
- Reliability

Training model process

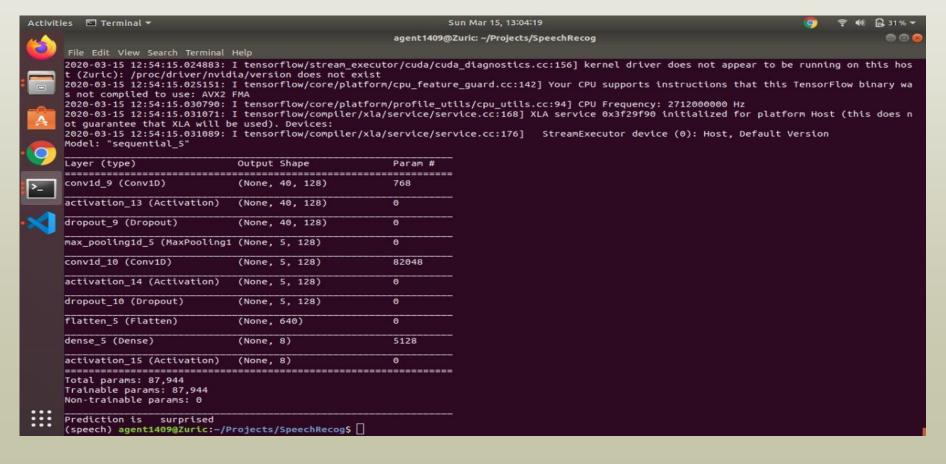


14 Use Case Model



Results

Snapshots



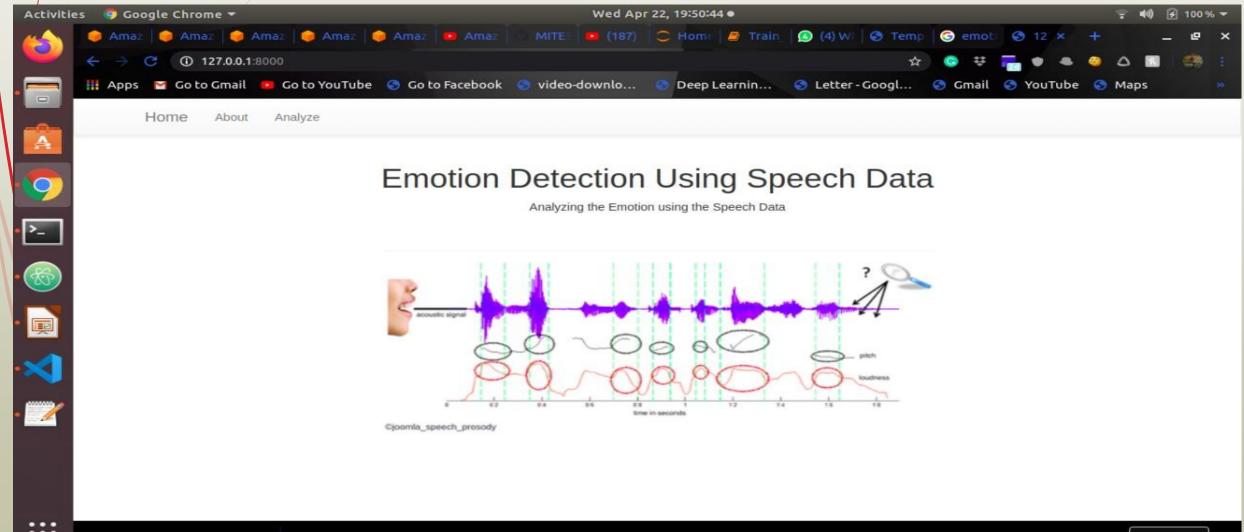
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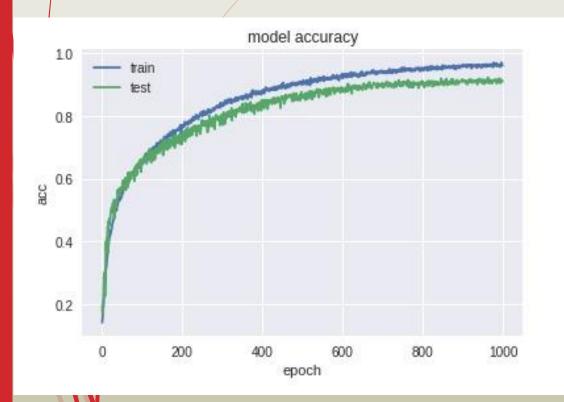
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6	File Edit View Search Terminal		agent1409@Zuric: ~/Proje		000
	t (Zuric): /proc/driver/nvid 2020-03-15 12:50:39.916210: s not compiled to use: AVX2 2020-03-15 12:50:39.921699: 2020-03-15 12:50:39.921896: ot guarantee that XLA will b	ia/version does not e I tensorflow/core/pla FMA I tensorflow/core/pla I tensorflow/compile e used). Devices:	exist atform/cpu_feature_guard.cc atform/profile_utils/cpu_ut r/xla/service/service.cc:16	cs.cc:156] kernel driver does not appear to be run 142] Your CPU supports instructions that this Tens ls.cc:94] CPU Frequency: 2712000000 Hz l] XLA service 0x5739a60 initialized for platform H s] StreamExecutor device (0): Host, Default Versi	sorFlow binary wa
>_	conv1d_9 (Conv1D)	(None, 40, 128)	768		
=	activation_13 (Activation)	(None, 40, 128)	0		
X	dropout_9 (Dropout)	(None, 40, 128)	0		
	max_pooling1d_5 (MaxPooling1	(None, 5, 128)	0		
	conv1d_10 (Conv1D)	(None, 5, 128)	82048		
	activation_14 (Activation)	(None, 5, 128)	0		
	dropout_10 (Dropout)	(None, 5, 128)	0		
	flatten_5 (Flatten)	(None, 640)	0		
	dense_5 (Dense)	(None, 8)	5128		
	activation_15 (Activation)	(None, 8)	0		
	Total params: 87,944 Trainable params: 87,944 Non-trainable params: 0				
:::	Prediction is neutral (speech) agent1409@Zuric:~/P	rojects/SpeechRecog\$	python3 livePredictions.py		i i

Activit	ies 🖸 Terminal ▼		Sun Mar 15, 13:0	4:27	
			agent1409@Zuric: ~/Project	s/SpeechRecog	00
	File Edit View Search Terminal I		. , , .		
	t (Zuric): /proc/driver/nvid 2020-03-15 12:53:40.812520: s not compiled to use: AVX2 2020-03-15 12:53:40.818111: ot guarantee that XLA will b 2020-03-15 12:53:40.818128:	ia/version does not I tensorflow/core/p FMA I tensorflow/core/p I tensorflow/compil e used). Devices:	exist latform/cpu_feature_guard.cc:1 latform/profile_utils/cpu_util er/xla/service/service.cc:168]	s.cc:156] kernel driver does not ap 42] Your CPU supports instructions s.cc:94] CPU Frequency: 2712000000 XLA service 0x44996c0 initialized StreamExecutor device (0): Host,	that this TensorFlow binary wa Hz for platform Host (this does n
0	Model: "sequential_5"				
	Layer (type)	Output Shape	Param #		
>_	conv1d_9 (Conv1D)	(None, 40, 128)	768		
	activation_13 (Activation)	(None, 40, 128)	0		
X	dropout_9 (Dropout)	(None, 40, 128)	0		
	max_pooling1d_5 (MaxPooling1	(None, 5, 128)	0		
	conv1d_10 (Conv1D)	(None, 5, 128)	82048		
	activation_14 (Activation)	(None, 5, 128)	0		
	dropout_10 (Dropout)	(None, 5, 128)	0		
	flatten_5 (Flatten)	(None, 640)	0		
	dense_5 (Dense)	(None, 8)	5128		
	activation_15 (Activation)	(None, 8)	0		
	Total params: 87,944 Trainable params: 87,944 Non-trainable params: 0				
:::	Prediction is sad (speech) agent1409@Zuric:~/P	roiects/SpeechRecog	S python3 livePredictions.py		

Home Page

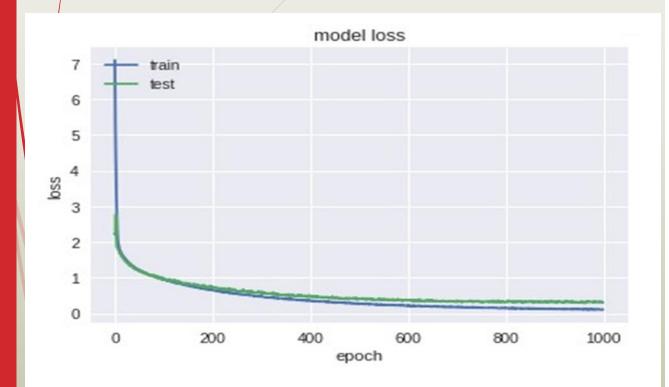


Accuracy graph



	precision	recall	fl-score	support
0	0.93	0.91	0.92	134
1	0.92	0.93	0.92	251
2	0.91	0.89	0.90	242
3	0.84	0.90	0.87	271
4	0.96	0.94	0.95	253
5	0.92	0.91	0.91	239
6	0.95	0.93	0.94	127
7	0.90	0.85	0.88	116
accuracy			0.91	1633
macro avg	0.92	0.91	0.91	1633
weighted avg	0.91	0.91	0.91	1633

```
from sklearn.metrics import confusion matrix
matrix = confusion matrix(new Ytest, predictions)
print (matrix)
# 0 = neutral, 1 = calm, 2 = happy, 3 = sad, 4 = angry, 5 = fearful, 6 = disgust, 7 = surprised
```



Layer (type)	Output	Shape	Param #
convld_3 (ConvlD)	(None,	40, 128)	768
activation_4 (Activation)	(None,	40, 128)	0
dropout_3 (Dropout)	(None,	40, 128)	0
max_pooling1d_2 (MaxPooling1	(None,	5, 128)	0
convld_4 (ConvlD)	(None,	5, 128)	82048
activation_5 (Activation)	(None,	5, 128)	0
dropout_4 (Dropout)	(None,	5, 128)	0
flatten_2 (Flatten)	(None,	640)	0
dense_2 (Dense)	(None,	8)	5128
activation_6 (Activation)	(None,	8)	0

Trainable params: 87,944 Non-trainable params: 0

Hardware and Software Requirement

SOFTWARE REQUIREMENTS

- ☐ Operating system : Windows 7 & higher
- ☐ Coding Language : Python 3 (LibROSA)
- ☐ Tool : Anaconda

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HARDWARE REQUIREMENTS

- ☐ System: Pentium Dual Core.
- ☐ Input Devices : Keyboard, Mouse
- ☐ Ram: 8 GB

Conclusion

- We have Obtained a Voice analysing Model which not only determines the words in the Sentence but also determines the emotions in them.
- Energy and pitch are two important features for voice emotion recognition.

References

- Andy Sun, Maisy Wieman ,Analysing Vocal Patterns to Determine Emotions 2018
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- Jangsik Cho, Shobei Kato, Detecting Emotions from Voice, 2011

THANK YOU