



# Speech Emotion Detection using IoT based Deep Learning

**CSE496 – Graduation Project  
Final Presentation**

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# Project Diagram and Description

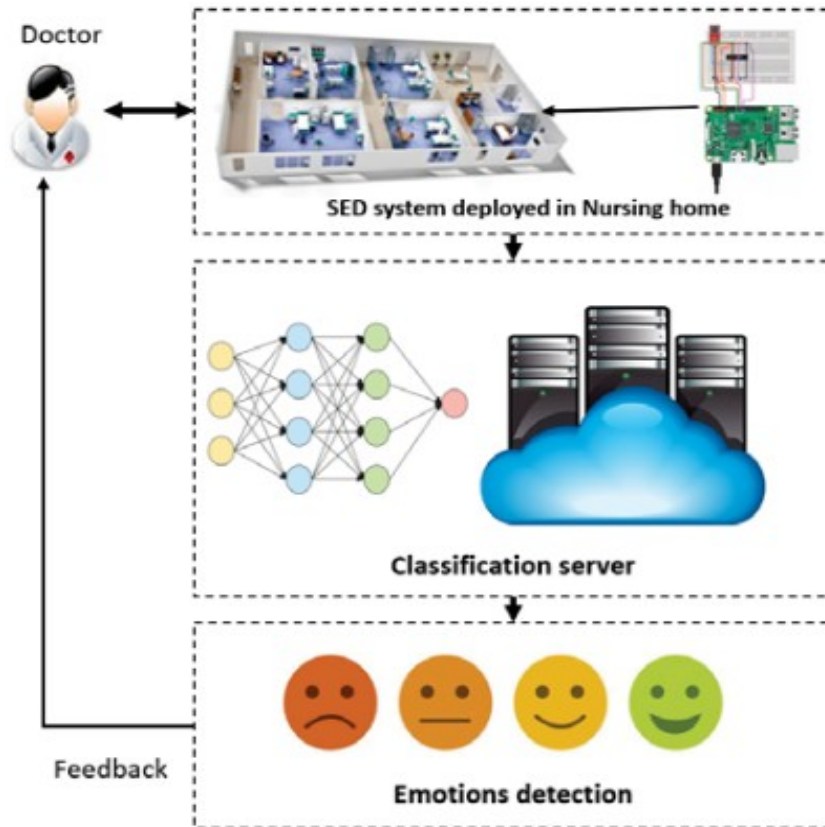
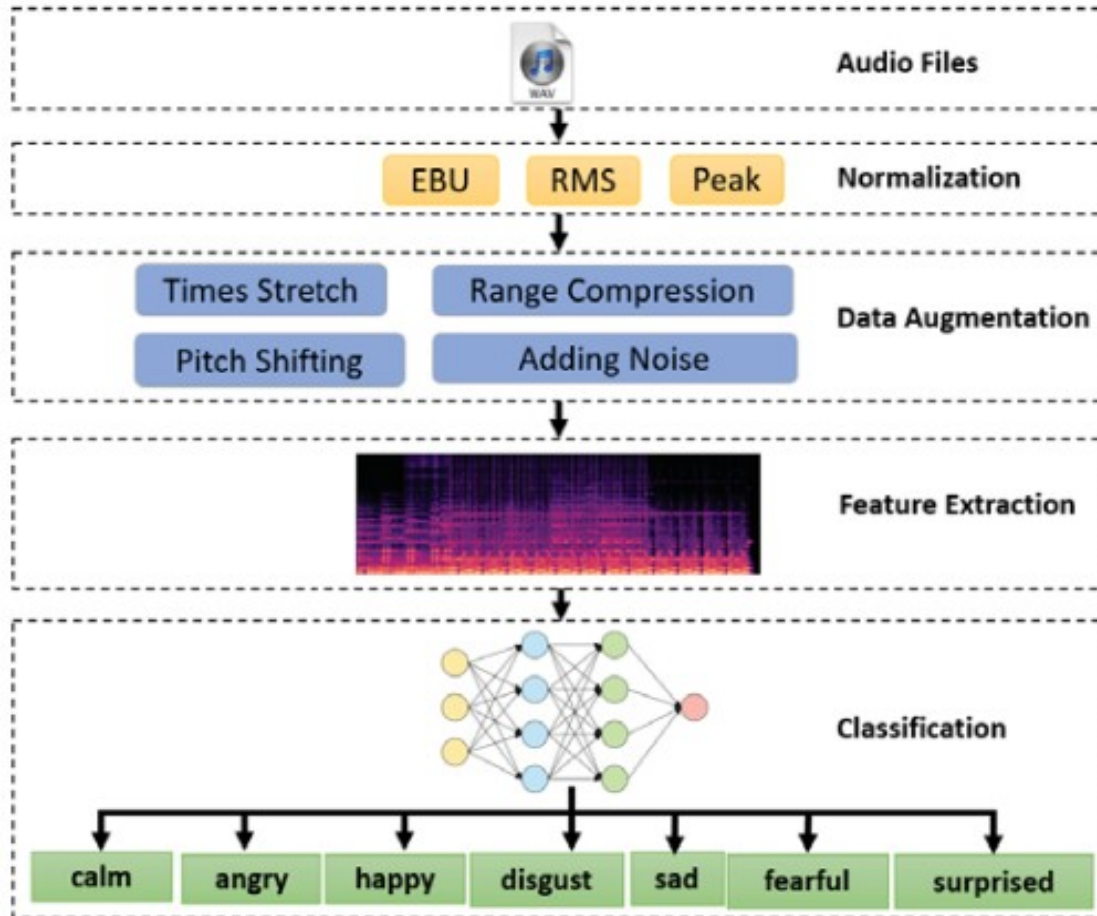


Fig. 1: The SED System: Real Time Speech Emotion Detection System based on Internet of Things (IoT) and Deep Learning for Health Care [1]

- Speech emotion detection system (SED)
- IoT device(mic) takes human voice as input
- Recognizes the human emotion
- Puts the recognition results on an interface
- Can be effective in many commercial applications



# Project Design Plan

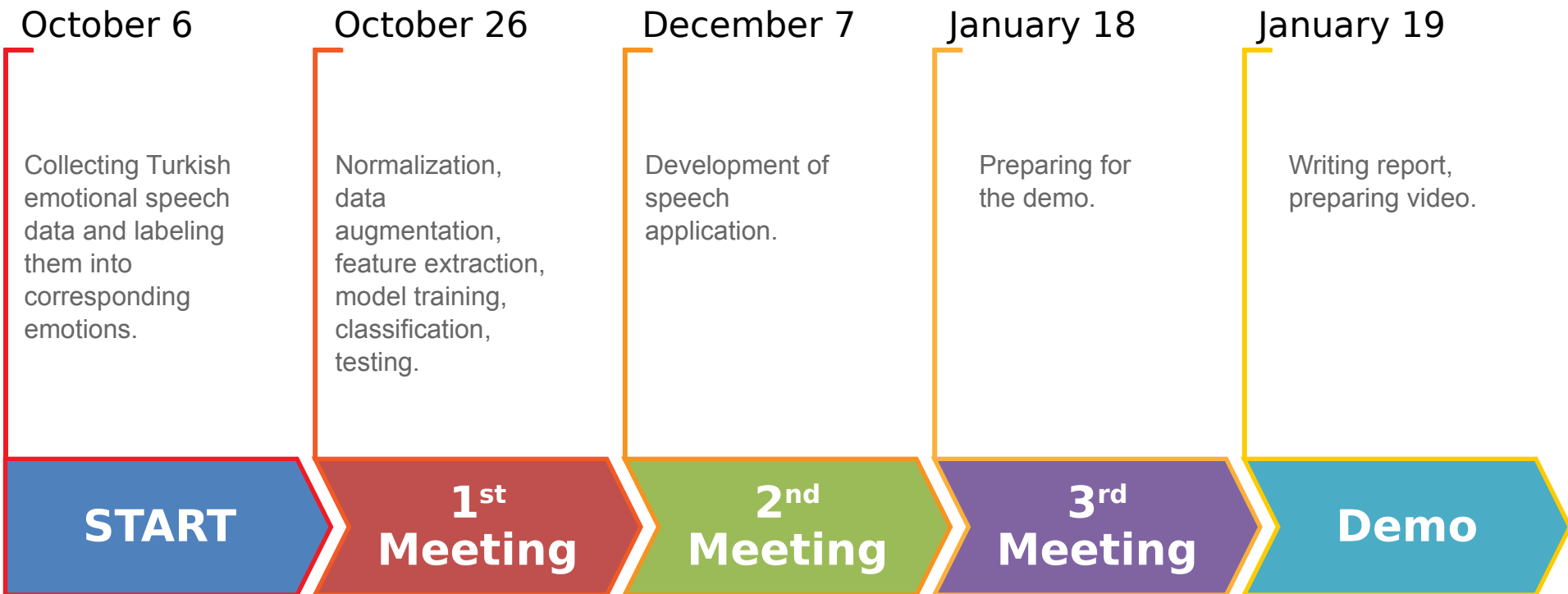


- Take audio as input
- Normalization
- Augmentation
- Feature extraction
- Classification

Fig. 2: The SED System: Classification Workflow for Speech Emotions [1]



# Project Timeline



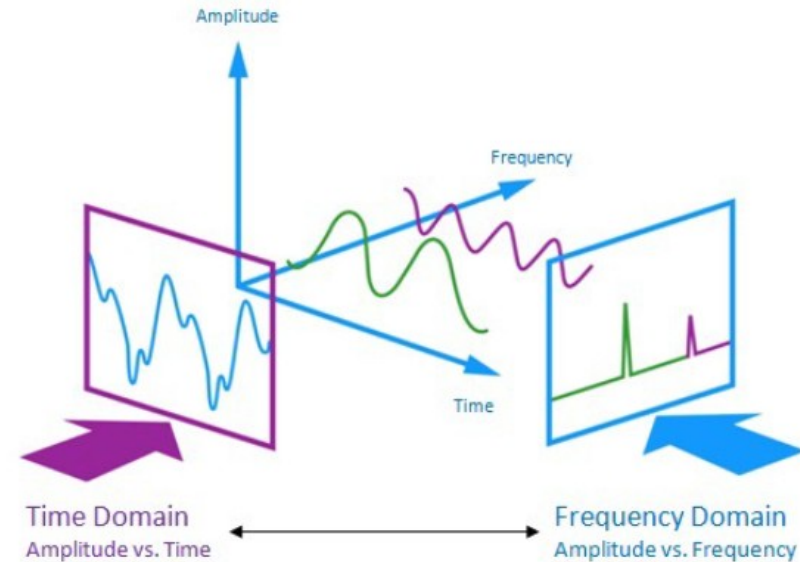
# Data Collection & Augmentation

- Turkish emotion voice database (TurEV-DB) [2]
  - Angry(487), Calm(408), Happy(357), Sad(483), total 1735 samples.
- Data collected from variable sources.
  - Radio theaters
  - Series/movies
  - Podcasts
- Data Augmentation
- In total we have 5418 Turkish emotional sample.



# Feature Extraction

- Important step
- A three-dimensional signal
- Sample rate and sample data



In this project I am only extracting 5 features:

- Zero Crossing Rate
- Chroma\_stft
- MFCC
- RMS(Root Mean Square) value
- MelSpectrogram



- CNN architecture
- MaxPooling
- Activation functions
  - ReLU
  - Softmax
- Adam optimizer

Model: "sequential"

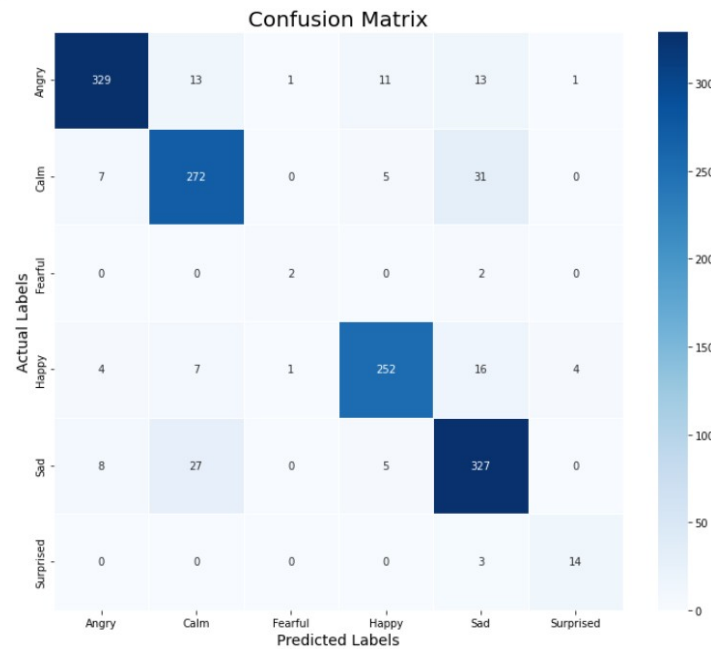
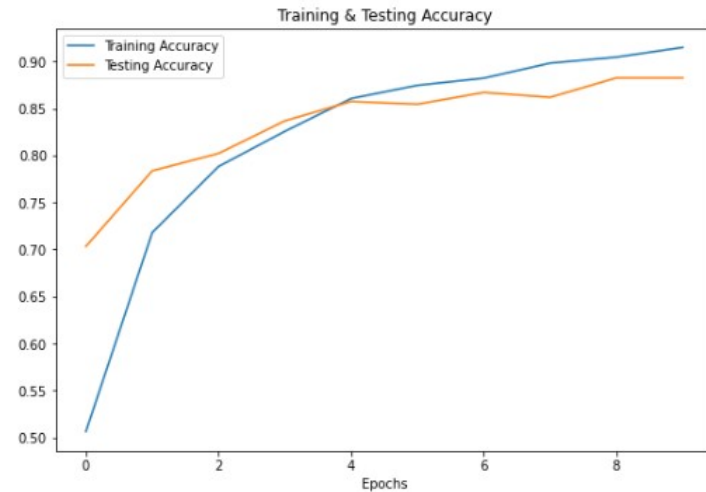
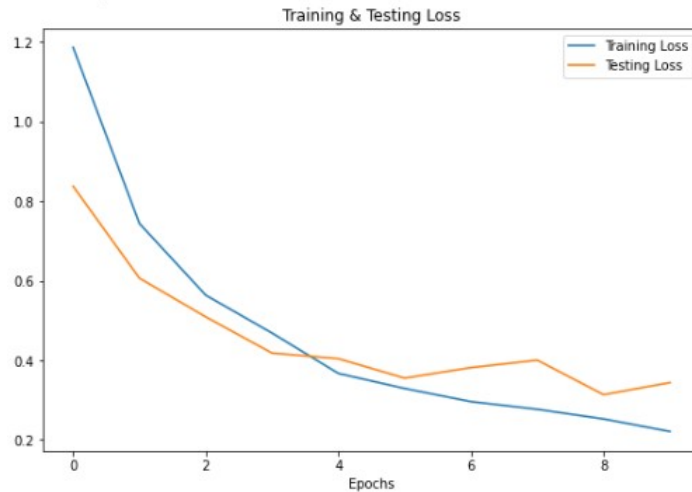
Layer (type)	Output Shape	Param #
conv1d (Conv1D)	(None, 162, 256)	1536
conv1d_1 (Conv1D)	(None, 162, 256)	327936
conv1d_2 (Conv1D)	(None, 162, 128)	163968
dropout (Dropout)	(None, 162, 128)	0
conv1d_3 (Conv1D)	(None, 162, 64)	41024
flatten (Flatten)	(None, 10368)	0
dense (Dense)	(None, 32)	331808
dropout_1 (Dropout)	(None, 32)	0
dense_1 (Dense)	(None, 6)	198
Total params: 866,470		
Trainable params: 866,470		
Non-trainable params: 0		



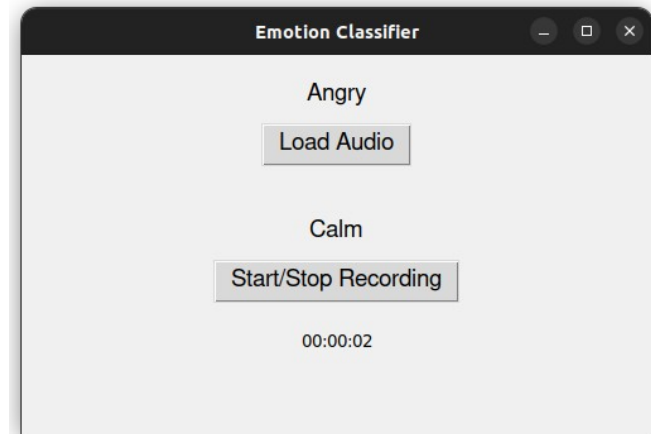
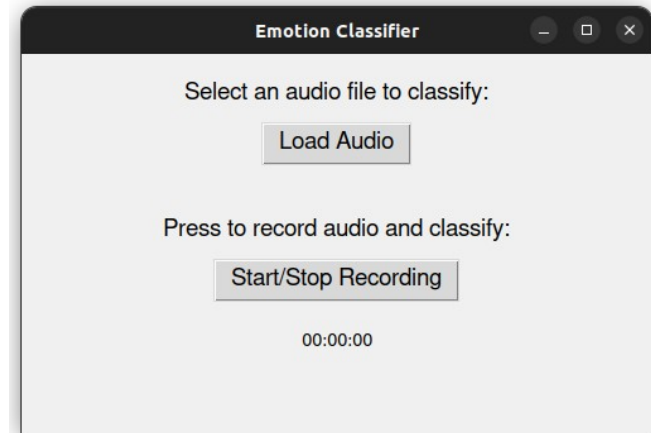
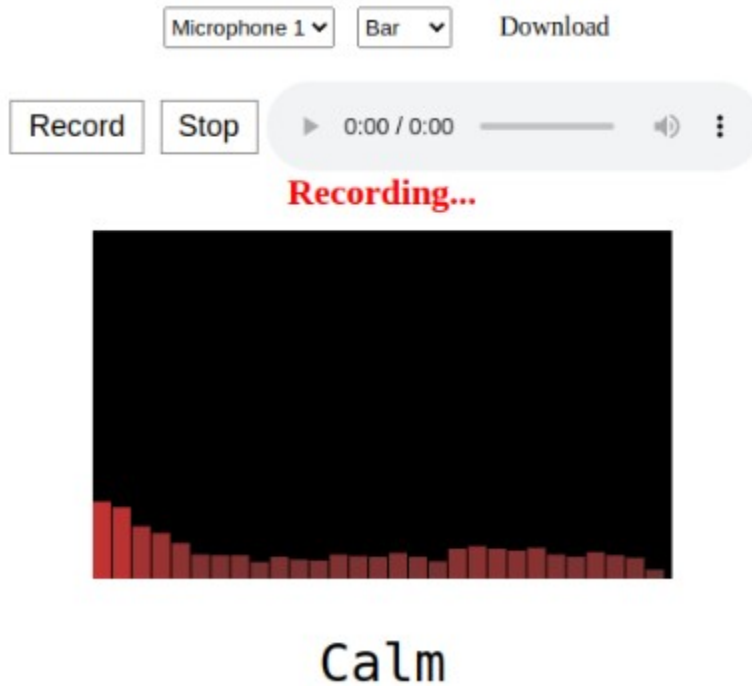


# Tests & Results

Accuracy of our model on test data : 88.26568126678467 %



# Application



- 1) Turkish emotion detection with at least a 90% accuracy rate X - 88% accuracy
- 2) Emotion detection in real-time ✓
- 3) At least ten thousand data will be used (Collected data + augmented data) X – 5418
- 4) The model is capable of detecting seven different emotions (calm, happy, sad, angry, fearful, surprised, and disgusted) X – Six, no disgusted



- Successfully developed a deep learning model for speech emotion detection
- Achieved 88% average accuracy
- Sample applications are developed
- Can be applied in call centers and in-car systems
- Limitations: Lack of data and samples
- Further improvements and data collection can increase accuracy
- A promising approach for understanding emotions in speech.



1. Tariq, Z., Shah, S.K. and Lee, Y., 2019, December. Speech emotion detection using iot based deep learning for health care. In 2019 IEEE International Conference on Big Data (Big Data) (pp. 4191-4196). IEEE.
2. S. F. Canpolat, Z. Ormanoğlu, and D. Zeyrek, “Turkish emotion voice database (turev-db),” in Proceedings of the 1st Joint Workshop on Spoken Language Technologies for Underresourced languages (SLTU) and Collaboration and Computing for UnderResourced Languages (CCURL), 2020, pp. 368–375.
3. N. Davis and K. Suresh, “Environmental sound classification using deep convolutional neural networks and data augmentation,” in 2018 IEEE Recent Advances in Intelligent Computational Systems (RAICS). IEEE, 2018, pp. 41–45.
4. Latif, S., Rana, R., Younis, S., Qadir, J. and Epps, J., 2018. Transfer learning for improving speech emotion classification accuracy. arXiv preprint arXiv:1801.06353.

