

# **Text Classification and Information Retrieval**

## **Automatic Recognition of Emotions from Recorded Speech Data**

Final Project
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## Introduction

Why Classification of Emotions'



#### **Emotions:**

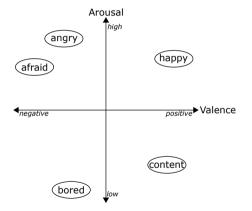


Figure: A two-dimensional emotion space with an Arousal and a Valence axis. Basic emotions are marked as ellipses within the quadrant.

### **Different Databases to classify emotions on:**



Figure: Types of databases used for emotion recognition and their difficulty.



### **Automatic Speech Emotion Recognition System:**

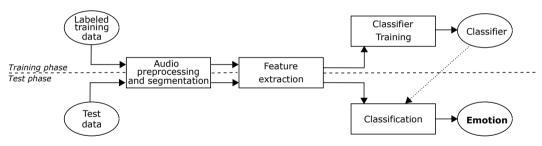


Figure: Overview of speech emotion recognition system.

## How did you feel

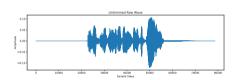


Figure: Un-trimmed sound wave of woman saying 'Dogs are sitting by the door.' in an fearful manner. There is still a lot of meaningless silence before and after the woman speaking.

### Dataset

- + Ryerson Audio-Visual Database of Emotional Speech and Song (RAVDESS)
- The RAVDESS database 'is a validated multimodal database of speech and song
- 24 professional actors and actresses, equally balanced in gender, were recorded phrasing main clauses such as 'Dogs are sitting by the door.'
- North American English
- in one of the following eight emotions: neutral, calm, happy, sad, angry, fearful, disgust, surprised.
- + 1440 RAVDESS files used (audio-only)

# Text Preprocssing



Turning speech into numbers

**Audio Preprocessing ...** 

**Speech Processing:** 

# Feature Extraction



...

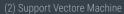
# Text Classification Algorithms

(1) Convolutional Neural Networl



- + CNNs are a special kind of neural network
- inspired by the concept of the mammalian retina
- have proven to perform well on high-dimensional input
- use convolution as a specialized kind of linear operation in place of a general matrix multiplication
- Advantageous to CNNs is their high efficiency in terms of computational complexity while using a sparse set of parameters
- The kernels which the CNN learns during training process, are reused over the entire input which markedly exceeds regular feed forward networks in terms of memory and computational efficiency.

# Classification Algorithms





### **Support Vector machine:**

- + **Goal:** find a hyperplane of the form  $\mathcal{H}=\left\{x\in\mathbb{R}^d|\left\langle w,x\right\rangle+b=0\right\}$  such that it separates the data while maximizing the distance between the hyperplane and the closest data point.
- + This distance is referred to as the margin.
- + By maximizing the margin, the classifier is most robust to noise in new data

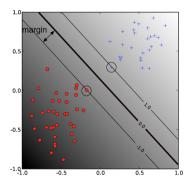


Figure: A linear SVM. The circled data points are the *support vectors* – i.e., the examples that are closest to the decision boundary. The support vectors determine the the margin with which the two classes are separated. In this work, eight classes need to be separated.



#### **SVM-Model Parameters:**

l = 745	(number of principal components of PCA
$\gamma_{ t PCA}=1$	(inversed kernel width of PCA)
C = 10	(soft-margin constant of SVM)
$\gamma_{\scriptscriptstyle \mathrm{SVM}}=10$	(inversed kernel width of SVM)

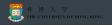
## **Comparison**

SVM	Training	Test
Accuracy	97.05%	63.39%

Table: Results of evaluating the SVM-model on the training and test data.



# Summary



#### SVM:

 $\begin{array}{lll} \text{SVM} & \text{Training} & \text{Test} \\ \text{Accuracy} & 97.05\% & 63.39\% \\ \end{array}$ 

### CNN:

- + did not learn the problem at all
- + probably training data too small
- +  $\rightarrow$  was not further pursued

### Improvements:

- incorporate transcribed text data
- + include videos to have facial expressions as additional feature
- using the validation set to find better hyperparameters
- +  $\rightarrow$  would make a plethora of established methods used in text processing available.
  - + feature extracting with BoW- or tf-idf method
  - + n-grams to capture context of a word

# Thank You!