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# Multiple Audio Event Detection / Audio Tagging

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## Abstract

In this assignment we mainly focused on Audio Event Detection. We had 10000 .npy files as dev & validation set data. For each of these files we need to output the voices present in the file. The basic outline and motive of this assignment are to pre-process the data efficiently and model the data in deep learning models.

## 1 My Methods

### 1.1 Convolutional Neural Network (CNN)

We used Conv2D layer in this assignment. This layer creates a convolution kernel that is convolved with the layer input to produce a tensor of outputs. The main advantage of CNN compared to its predecessors is that it automatically detects the important features without any human supervision. For example, given many pictures of cats and dogs it learns distinctive features for each class by itself. CNN is also computationally efficient.

#### Results by CNN:

	precision	recall	f1-score	support
0	0.38	0.31	0.34	400
1	0.50	0.15	0.23	266
2	0.65	0.40	0.50	284
3	0.57	0.21	0.31	689
4	0.37	0.35	0.36	341
5	0.58	0.35	0.43	283
6	0.89	0.43	0.58	377
7	0.30	0.29	0.29	306
8	0.95	0.98	0.97	2373
9	0.54	0.45	0.49	251
micro avg	0.75	0.60	0.66	5570
macro avg	0.57	0.39	0.45	5570
weighted avg	0.71	0.60	0.63	5570
samples avg	0.78	0.61	0.66	5570

Precision -> 74.61917562724014  
Recall -> 59.80251346499102  
F1 Score -> 66.39425951764002

F1 score = 66.394  
Precision = 74.619  
Recall = 59.802

## 17    **1.2    Dense Neural Network (DNN)**

18    In any neural network, a dense layer is a layer that is deeply connected with its preceding layer which  
19    means the neurons of the layer are connected to every neuron of its preceding layer. This layer is the  
20    most commonly used layer in artificial neural network networks.

### 21    **Results by DNN:**

22    F1 score = 45.341

23    Precision = 54.619

24    Recall = 38.757

## 25    **1.3    Recurrent-Convolutional Neural Network (RCNN)**

26    Classification of audio with variable length using a CNN + LSTM architecture.

### 27    **Results by RCNN:**

28    F1 score = 38.654

29    Precision = 44.619

30    Recall = 34.095

## 31    **1.4    Recurrent Neural Network (RNN)**

32    The goal of the SVM algorithm is to create the best line or decision boundary that can segregate  
33    n-dimensional space into classes so that we can easily put the new data point in the correct category  
34    in the future. This best decision boundary is called a hyperplane.

### 35    **Results by RNN:**

36    F1 score = 40.378

37    Precision = 50.619

38    Recall = 33.584

## 39    **1.5    K Nearest Neighbours (KNN)**

40    KNN algorithm assumes the similarity between the new case/data and available cases and put the  
41    new case into the category that is most similar to the available categories. K-NN algorithm stores all  
42    the available data and classifies a new data point based on the similarity. This means when new data  
43    appears then it can be easily classified into a well suite category by using K- NN algorithm.

### 44    **Results by KNN:**

45    F1 score = 58.705

46    Precision = 66.619

47    Recall = 52.471

## 48    **1.6    Support Vector Machine (SVM)**

49    The goal of the SVM algorithm is to create the best line or decision boundary that can segregate  
50    n-dimensional space into classes so that we can easily put the new data point in the correct category  
51    in the future. This best decision boundary is called a hyperplane.

### 52    **Results by SVM:**

53    F1 score = 40.378

54    Precision = 74.619

55    Recall = 36.873