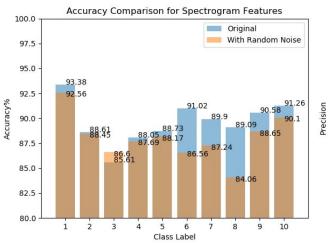
MCA Assignment - 2

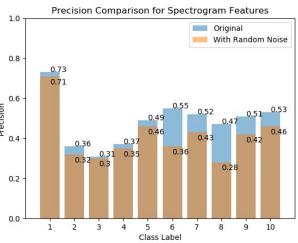
Arushi Chauhan 2016019

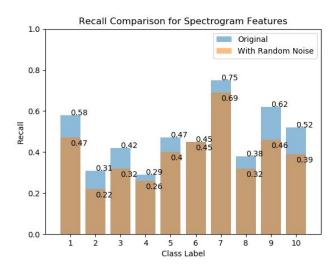
Spectrogram Features

Results

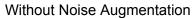
Accuracy: 40.25% (with noise augmentation), 48.11% (without noise augmentation)



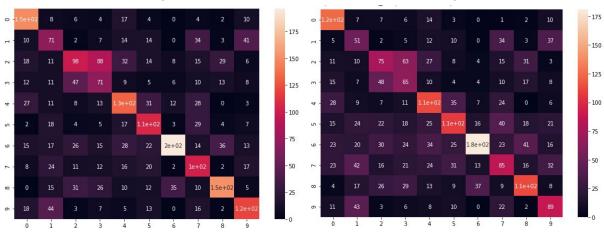




Confusion Matrices



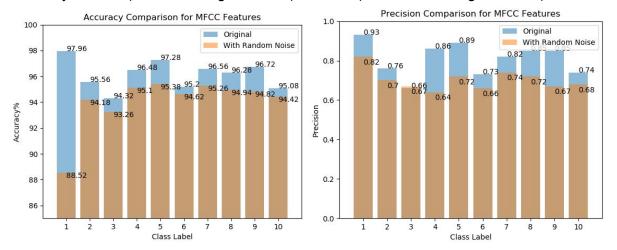
With Noise Augmentation

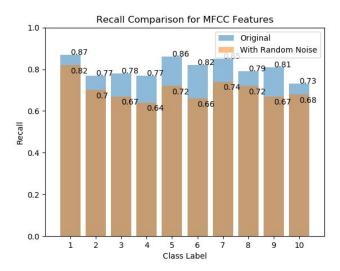


MFCC Features

Results

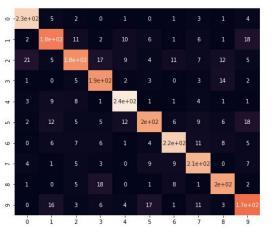
Accuracy: 70.28%(with noise augmentation), 80.67% (without noise augmentation)



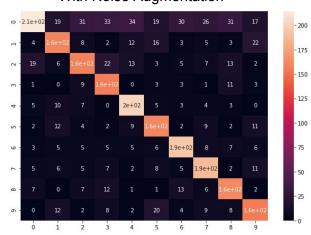


Confusion Matrices

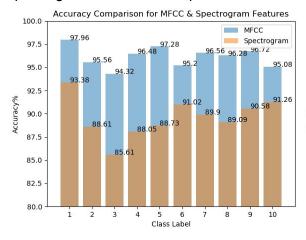
Without Noise Augmentation

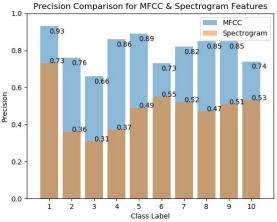


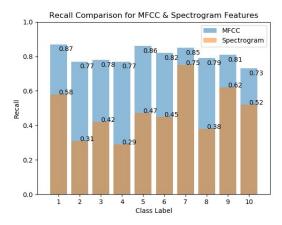
With Noise Augmentation



Spectrogram vs MFCC Comparison







Noise Augmentation Process

Noise was added to 30% of the training and validation samples randomly. Since there were six noise samples, one of them was chosen randomly for a given training/validation sample. The resultant wave was a 60:40 combination of original sound and chosen noise sample.

Observations:

- MFCC consistently outperforms spectrogram in terms of higher accuracy, precision and recall for all classes. Hence, it can be considered as a better feature in comparison to spectrogram.
- 2. Models performed better without noise augmentation in terms of higher accuracy, precision and recall reported in original models as compared to models trained on noise augmented data.

Spectrogram Plots

