

# **Speech Command Recognition**

Made By:

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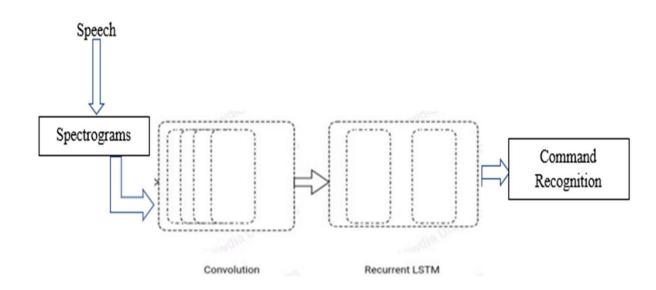
## What is Speech Command Recognition?



Speech command recognition has extensive real-world applications to identify the import words/phrases in the spoken sentences.

One such application is Keyword spotting (KWS) technology, a potential technique to provide fully hands-free interface, and this is especially convenient for mobile devices compared to typing by hands.

Implementation of keyword spotting technology involves networks which helps in understanding the specific commands in the conversations.



Source: (PDF) A Hybrid Technique using CNN+LSTM for Speech Emotion Recognition (researchgate.net)

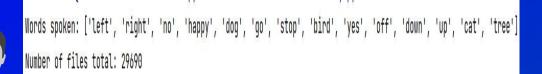
### **Dataset Description**





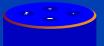
The dataset used in this project is sourced from Kaggle which has audio files (.wav) of duration of one second. The dataset has approximately 30K audio files classified based on 14 major labels. The dataset will be separated into train-test sets manually and processed to be trained on the deep learning models.

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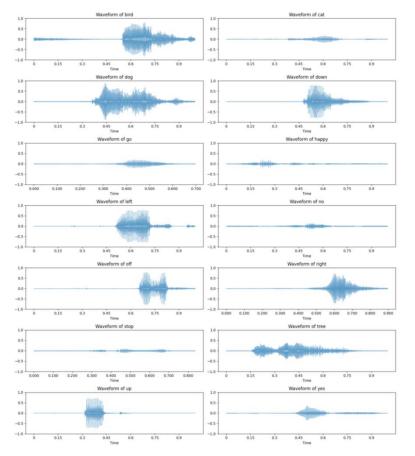


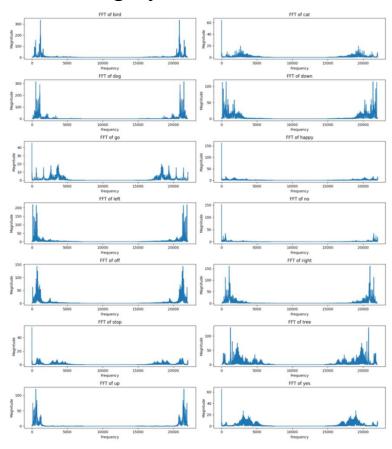
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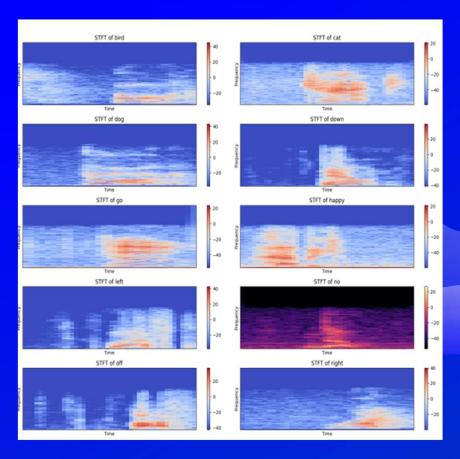
```
data = {
    "mapping": [],
    "labels": [],
    "MFCCs": [],
    "files": []
}
```



### Waveforms and FFT of each Sub Category

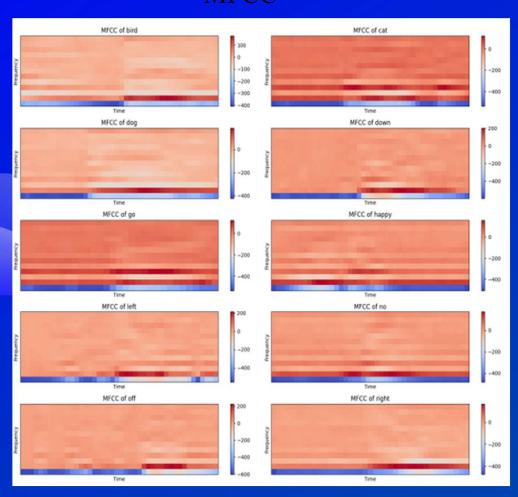






STFT

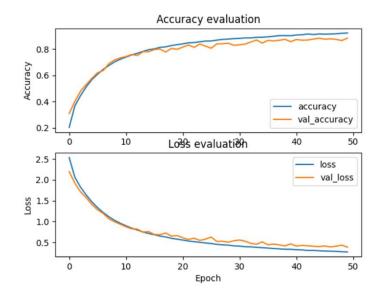
### MFCC



#### **CNN**

CNN model on training on 50 epochs resulted in accuracy of 87% on the test set. The accuracy and loss evaluation plots shows that the model has been training better on the train data as we can see a smooth curve. When it was validated based on validation set there has been a fluctuations in the loss and accuracy values which shows the model might be overseeing the data.

Test loss: 0.39549198746681213, test accuracy: 87.9107117652893



### CNN + LSTM



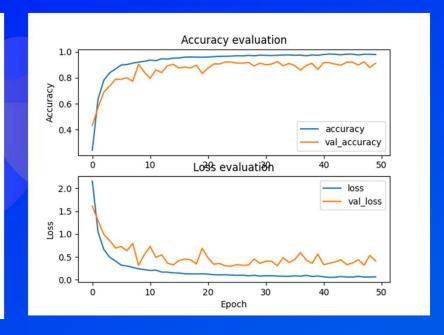
A convolutional neural network being a feed-forward network, filters spatial data whereas the recurrent neural network (LSTM) feeds data back into itself. Thus, recurrent neural networks are better suited for sequential data. A convolutional neural network can perceive patterns across space, LSTM can see them over time. Since speech signal is sequential, so LSTM is best suited for speech processing.

```
204/204 [========] - 2s 12ms/step - loss: 0.0768 - accuracy: 0.9746 - val_loss: 0.4394 - val_accuracy: 0.8972
Epoch 48/50
204/204 [========] - 2s 12ms/step - loss: 0.0602 - accuracy: 0.9803 - val_loss: 0.3220 - val_accuracy: 0.9223
Epoch 49/50
204/204 [==========] - 2s 12ms/step - loss: 0.0608 - accuracy: 0.9798 - val_loss: 0.5318 - val_accuracy: 0.8780
Epoch 50/50
204/204 [===========] - 2s 12ms/step - loss: 0.0650 - accuracy: 0.9779 - val_loss: 0.4105 - val_accuracy: 0.9115

259/250 [==========] - 1s 3ms/step - loss: 0.4082 - accuracy: 0.9057

Test loss: 0.4082329571247101, test accuracy: 90.5693531036377

Process finished with exit code 0
```



### CNN + LSTM

conv2d (Conv2D)	(None, 44, 13, 128)	1280		
batch_normalization (BatchN ormalization)	(None, 44, 13, 128)	512		
activation (Activation)	(None, 44, 13, 128)	0		
<pre>max_pooling2d (MaxPooling2D )</pre>	(None, 22, 7, 128)	0		
conv2d_1 (Conv2D)	(None, 22, 7, 72)	83016		
batch_normalization_1 (Batc hNormalization)	(None, 22, 7, 72)	288		
activation_1 (Activation)	(None, 22, 7, 72)	0		
max_pooling2d_1 (MaxPooling 2D)	(None, 6, 2, 72)	0		
conv2d_2 (Conv2D)	(None, 6, 2, 64)	41536		
batch_normalization_2 (Batc hNormalization)	(None, 6, 2, 64)	256		
activation_2 (Activation)	(None, 6, 2, 64)	0		
max_pooling2d_2 (MaxPooling 2D)	(None, 2, 1, 64)	0		
conv2d_3 (Conv2D)	(None, 2, 1, 64)	36928		
batch_normalization_3 (Batc hNormalization)	(None, 2, 1, 64)	256		
activation_3 (Activation)	(None, 2, 1, 64)	0		
max_pooling2d_3 (MaxPooling 2D)	(None, 1, 1, 64)	0		
reshape (Reshape)	(None, 1, 64)	0		
1stm (LSTM)	(None, 32)	12416		
dense (Dense)	(None, 14)	462		

# Results

Model	Train accuracy	Train loss	Test accuracy	Test loss
CNN	0.91	0.30	0.87	0.39
CNN+LSTM	0.97	0.06	0.905	0.40

