

INTRODUCTION TO AI AND ML Toycar Presentation

P.Srijith Reddy,
EE19BTECH11041,
Dept. of Electrical Engg.,
IIT Hyderabad.

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INTRODUCTION
TO AI AND ML
Toycar
Presentation

P.Srijith Reddy,
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Engg.,
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Outline

ZEROPADDING

MFCC's

RNN

Gradient Descent
Method

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Zeropadding

Firstly we collect voice samples of required speech and then we pad them such that the recorded voice concentrates on the central part of padded samples without any noise added. we collected about 80 samples and made around 2000 samples of each speech.

Mel-frequency cepstral coefficients (MFCCs)

MFCC is a representation of the short-term power spectrum of a sound, which in simple terms represents the shape of the vocal tract. and when we load the data for training we compute mfcc's as vector of size [49,39].
and we label speech parts with some numbers.

$$Mel(f) = 1125 \ln\left(1 + \frac{f}{700}\right) \quad (3.1)$$

CODE:

```
back, sr = sf.read("Final/back"+str(i)+".wav")
back, index = librosa.effects.trim(back)
data.append(mfcc(y = back, sr = sr, nmfcc = 39).T)
label.append(0)
```

Recurrent neural network A recurrent neural network (RNN) is a class of artificial neural networks where connections between nodes form a directed graph along a temporal sequence. RNNs can use their internal state (memory) to process sequences of inputs.

LSTM Long Short Term Memory networks usually just called LSTMs.

LSTM's are special kind of RNN's which can remember memory for long periods.

CODE

```
model.add(LSTM(units = 128, return_sequences = True,  
input_shape = (data.shape[1],39)))
```

Gradient Descent Method

let's say x is input and y is original output and y' as output.
Here for example forward is represented as

$$y = [1, 0, 0, 0, 0]^T \text{ and}$$

similarly other parts. (5.1)

$$y' = \text{sigmoid}(W.X + B) \quad (5.2)$$

$$J(W, b) = 1/2(\sum (y - y')^2) \quad (5.3)$$

where

$$\text{sigmoid}(x) = (1/(1 + e^{-x})) \quad (5.4)$$

Here $J(W, b)$ measures how close is output to original output and we need to minimize it's value.

Loss function

$J(W, b)$ can be minimized using gradient descent method.
The loss function here is let's say E

$$E = - \sum_{i=0}^C y_i \log(y'_i) \quad (5.5)$$

(**CODE**

```
def categorical_cross_entropy(ytrue, ypred, axis=-1):  
    return -1.0*tf.reduce_mean(tf.reduce_sum(ytrue *  
        tf.log(ypred), axis))
```

Softmax function

$$f(s_i) = \frac{e^{s_i}}{\sum_j^C e^{s_j}} \quad (5.6)$$

This function is used to generate outputs in the range of (0,1). eventually whose probabilities sum to 1

*Thank
you!*

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