

# Deep speech

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**This document present our project in machine learning. We have implemented a voice recognition system i.e. our program is able to recognize spoken language and translate into text by computers. We use python3, Keras and Tensorflow.**

## Introduction

We have implemented this article (1) with python3, Keras and Tensorflow. The neuron network used is not common. So we created our own neuron network model.

## Model

We have seven layers of neuron. The three first layers are computed by :

$$h_t^{(l)} = g(W^{(l)}h_t^{(l-1)} + b^{(l)})$$

where  $g(z) = \min\{\max\{0, z\}, 20\}$  and  $W^{(l)}, b^{(l)}$  are the weight matrix and bias parameters for layers  $l$ .

The fourth layer is a bi-directional recurrent layer. This layer includes two sets of hidden units : a set with forward recurrence,  $h^{(f)}$ , and a set with backward recurrence  $h^{(b)}$  :

$$h_t^{(f)} = g(W^{(4)}h_t^{(3)} + W_r^{(f)}h_{t-1}^{(f)} + b^{(4)})$$

$$h_t^{(b)} = g(W^{(4)}h_t^{(3)} + W_r^{(b)}h_{t+1}^{(b)} + b^{(4)})$$

The fifth (non-recurrent) layer takes both the forward and backward units as inputs  $h_t^{(5)} = g(W^{(5)}h_t^{(4)} + b^{(5)})$  where  $h_t^{(4)} = h_t^{(f)} + h_t^{(b)}$ . The output layer is a standar softmax function that yields the predicted character probabilities for each time slice  $t$  and character  $k$  in the alphabet :

$$h_{t,k}^{(6)} \equiv \mathbb{P}(c_t = k|x) = \frac{\exp(W_k^{(6)}h_t^{(5)} + b_k^{(6)})}{\sum_j \exp(W_j^{(6)}h_t^{(5)} + b_j^{(6)})}$$

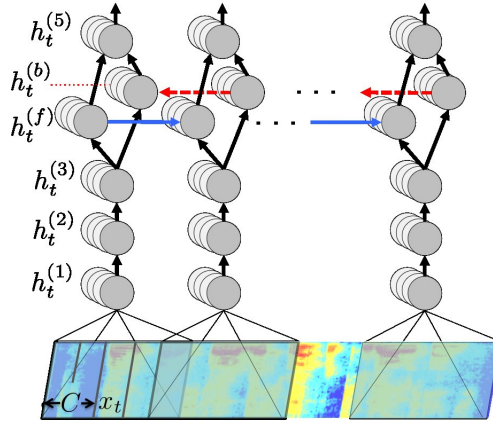


Figure 1: Structur of our RNN model

## Our work

This article (*1*) create a new model and use a ctc loss function. So to create this model, we have customized our model so that it is like on the article. For that we had to work a lot on the documentation of keras and tensorflow. However our main problem was the ctc loss function but finally everything is good.

## References and Notes

1. A. Y. Hannun, *et al.*, *CoRR* **abs/1412.5567** (2014).