PROJECT STATUS

Problem Statement:

Gender Recognition system from audio files using FFT with Artificial Neural Networks.

Status:

6. Training

Using Artifcial Neural Network to train the model for classification problem.

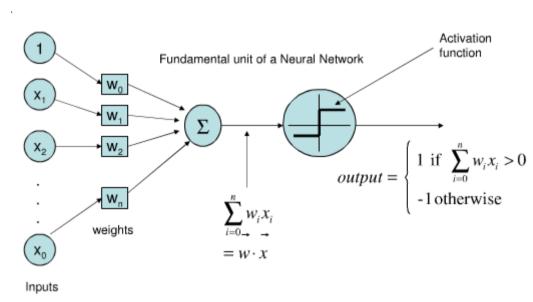


Illustration 1: Artificial Neural Network

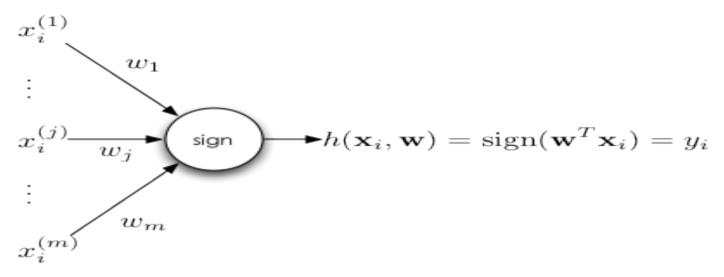
The description of the architecture i am using for classification are as follows:

- 1. 1 input layer, Hidden layer, 1 Output Layer.
- 2. using 12 input attributes and 1 output attributes.
- 3. activation function for input and hidden layer is "relu".
- 4. activation function for output layer is "sigmoid".
- 5. using 10 perceptron in each input and hidden layer.
- 6. for optimisation algorthm is used "adam".

Q. What is Perceptron?

Ans:

a computer model or computerized machine devised to represent or simulate the ability of the brain to recognize and discriminate.



Q.What is activation Function?

Ans:

In artificial neural networks, the activation function of a node defines the output of that node given an input or set of inputs. A standard computer chip circuit can be seen as a digital network of activation functions that can be "ON" (1) or "OFF" (0), depending on input.

ReLu activation Function:

A unit employing the rectifier is also called a rectified linear unit (ReLU). A smooth approximation to the rectifier is the analytic function

$$F(x) = Log(1 + exp(x))$$

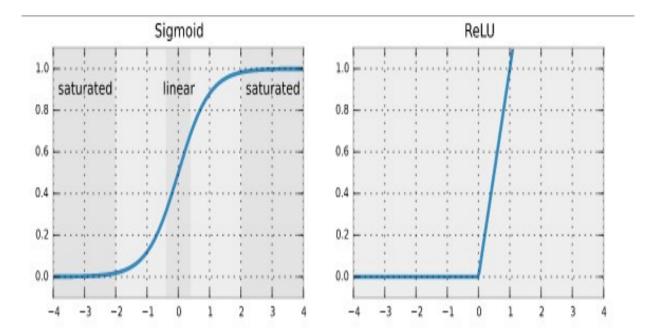
which is called the **softplus** function

Sigmoid activaton Function:

It is used in neural networks to give logistic neurons real-valued output that is a smooth and bounded function of their total input. It also has the added benefit of having nice derivatives which make learning the weights of a neural network easier.

$$F(x) = 1/(1 + exp(-x))$$

Graphical Representation



Implementation

Modules used:

import numpy as np
import pandas as pd
import h5py
from sklearn.model_selection import train_test_split
from keras.models import load_model
from keras.models import Sequential
from keras.layers import Dense, Dropout
from keras import optimizers
from sklearn.utils import shuffle
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix

Using python keras module for layer creation and compilation:

```
# Initialising the ANN
classifier = Sequential()

# Adding the input layer and the first hidden layer
classifier.add(Dense(output_dim = 10, init = 'uniform', activation = 'relu', input_dim = 12))

# Adding the second hidden layer
classifier.add(Dense(output_dim = 10, init = 'uniform', activation = 'relu'))

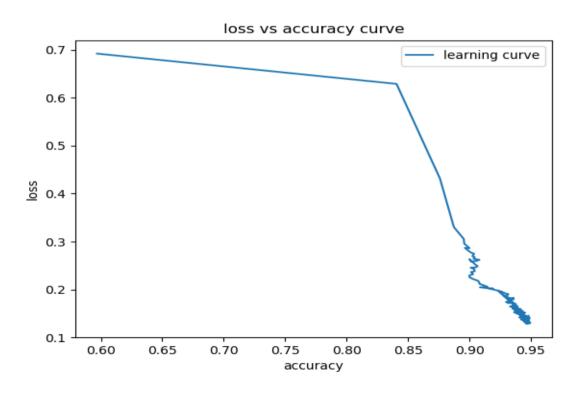
# Adding the third hidden layer
classifier.add(Dense(output_dim = 10, init = 'uniform', activation = 'relu'))

# Adding the output layer
classifier.add(Dense(output_dim = 1, init = 'uniform', activation = 'sigmoid'))

# Compiling the ANN
classifier.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])

# Fitting the ANN to the Training set
history = classifier.fit(x_train, y_train, batch_size = 20, nb_epoch = 200)
```

Learnng Curve:



Trainng process:

Epoch 1/200
2018-06-17 00:32:26.187508: I tensorflow/core/platform/cpu_feature_guard.cc:137] Your CPU
supports instructions that this TensorFlow binary was not compiled to use: SSE4.1 SSE4.2 AVX
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7. Testing and Evaluation

Q.What is Confusion Matrix?

Ans:

A confusion matrix is a table that is often used to describe the performance of a classification model (or "classifier") on a set of test data for which the true values are known.

For the above Model:

Accuracy : (correctly predicted class / total testing class) \times 100%

```
Training :- 94%
Testing :- 92.44%
```

Precision = TP / TP + FP

```
=489/(489+39)
```

= 0.9261363636363636

= 92.61%

Recall = TP/TP+FN

=489/(489+40)

= 0.9243856332703214

= 92.43%

F1-Score: = 2*(Recall * Precision) / (Recall + Precision)

```
=2*(0.9261363636363636*0.9243856332703214)/(0.926136363636363636+0.9243856332703214)
```

- = 0.925252443830496
- = 92.52%