

Robotics Inference

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Abstract - This paper discusses project that that uses NVIDIA DIGITS workflow to train two deep neural networks to perform inference. The first one classify bottles, candy wrappers and nothing on a moving belt and the training dataset was supplied by Udacity while the second network classify three Nigeria's banknotes which include 200, 500 and 1,000 Naira and it is student own idea and the dataset also collected by the student.

1.0 INTRODUCTION

The traditional method of optical sensor has been used detect and identify different banknotes, however, use of neural to classify banknote will offer more effective and flexible alternative. In this project GoogleNet neural network is utilized to train and classify three Nigeria's banknote denominations as well as supplied data by supplying thousands of sample images. This enable these networks learn important color, geometric and pattern information from the images and make prediction accordingly.

2.0 BACKGROUND/FORMULATION

This project is an image classification problem which requires supervised learning approach. Since supervised learning means predicting input values based on the label of the training examples that you have previously provided. Therefore, the GoogleNet is used because is a good model for solving image classification problems and it has a good inference per image with considerable accuracy. The parameters used are Stochastic Gradient Descend (SGD) with the learning of 0.01 and 30 epochs for bank note classification while P1 (supplied data) classification parameters are Stochastic Gradient Descend (SGD) with the learning of 0.001 and 5 epochs.

DIGITS
New Model
salab (Logout)
Info
About

New Image Classification Model

Select Dataset

banknote_inference

banknote_inference

Done 10:54:47 AM

Image Size
256x256

Image Type
COLOR

DB backend
Imdb

Create DB (train)
2364 images

Create DB (val)
797 images

Create DB (test)
32 images

Python Layers

Server-side file

Use client-side file

Solver Options

Training epochs
30

Snapshot interval (in epochs)
1

Validation interval (in epochs)
1

Random seed
[none]

Batch size
[network defaults] multiples allowed

Batch Accumulation

Solver type
SGD (Stochastic Gradient Descent)

Base Learning Rate
0.01 multiples allowed

Show advanced learning rate options

Data Transformations

Subtract Mean
Image

Crop Size
none

Activate W
Go to Action C

Figure 1: Banknote creation model Parameters

DIGITS
New Model
salab (Logout)
Info
About

New Image Classification Model

Select Dataset

supplied_data

supplied_data

Done 10:54:47 AM

Image Size
256x256

Image Type
COLOR

DB backend
Imdb

Create DB (train)
7170 images

Create DB (val)
2324 images

Python Layers

Server-side file

Use client-side file

Solver Options

Training epochs
5

Snapshot interval (in epochs)
1

Validation interval (in epochs)
1

Random seed
[none]

Batch size
[network defaults] multiples allowed

Batch Accumulation

Solver type
SGD (Stochastic Gradient Descent)

Base Learning Rate
0.001 multiples allowed

Show advanced learning rate options

Data Transformations

Subtract Mean
Image

Crop Size
none

Standard Networks	Previous Networks	Pretrained Networks	Custom Network												
<div>Caffe</div> <table border="1"> <thead> <tr> <th>Network</th> <th>Details</th> <th>Intended image size</th> </tr> </thead> <tbody> <tr> <td>LeNet</td> <td>Original paper [1998]</td> <td>28x28 (gray)</td> </tr> <tr> <td>AlexNet</td> <td>Original paper [2012]</td> <td>256x256</td> </tr> <tr> <td>GoogLeNet</td> <td>Original paper [2014]</td> <td>256x256</td> </tr> </tbody> </table>	Network	Details	Intended image size	LeNet	Original paper [1998]	28x28 (gray)	AlexNet	Original paper [2012]	256x256	GoogLeNet	Original paper [2014]	256x256			Customize
Network	Details	Intended image size													
LeNet	Original paper [1998]	28x28 (gray)													
AlexNet	Original paper [2012]	256x256													
GoogLeNet	Original paper [2014]	256x256													

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Figure 2: Supplied data Model creation parameters

3.0 DATA ACQUISITION

P1 dataset is provided by Udacity while banknote images dataset were taken with Infinix Hot 4 Pro Android phone and used Augmentor python library to generate

more additional images in order to have enough data for better network accuracy.

The images are RGB color and were resized to 256x256.

Below are details for the images collected for the network training:-

	200 note	500 note	1000 note
Phone Images	322	311	260
Augmented images	804	851	650
Total	1,126	1,162	910

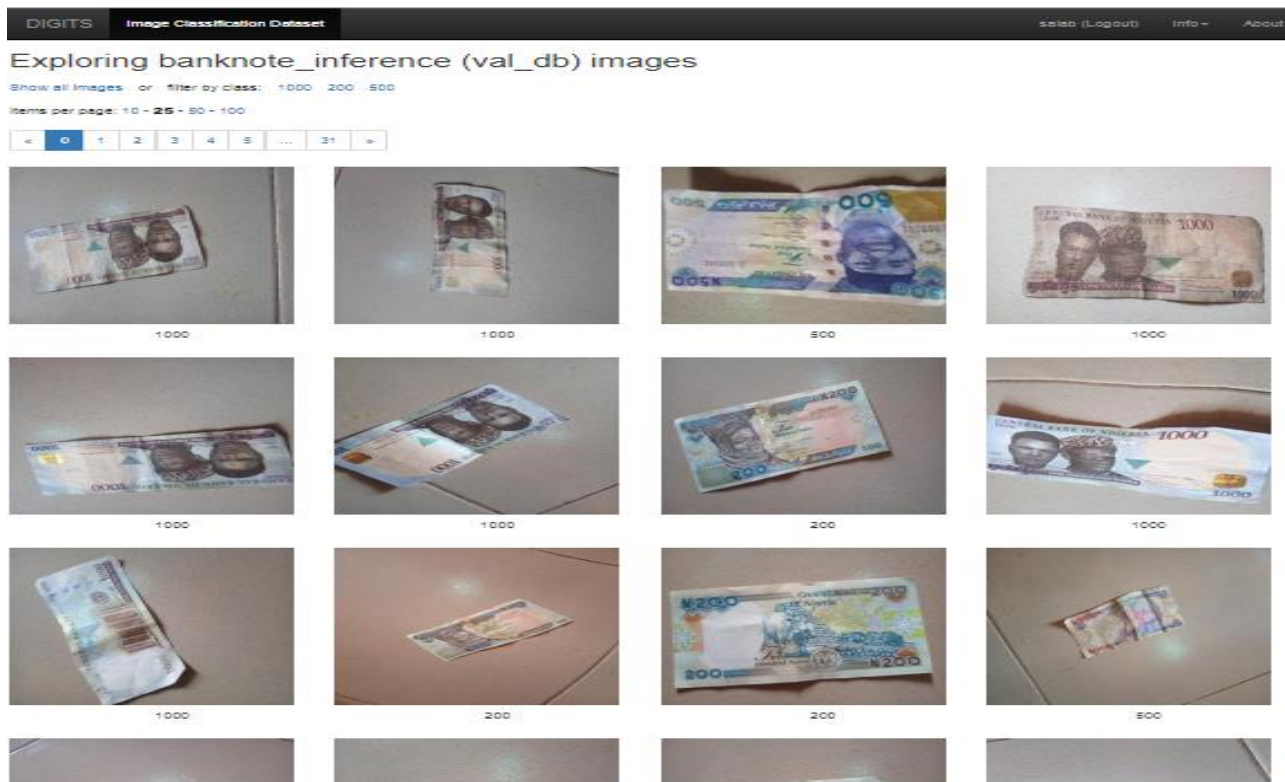


Figure 3: Banknote sample images

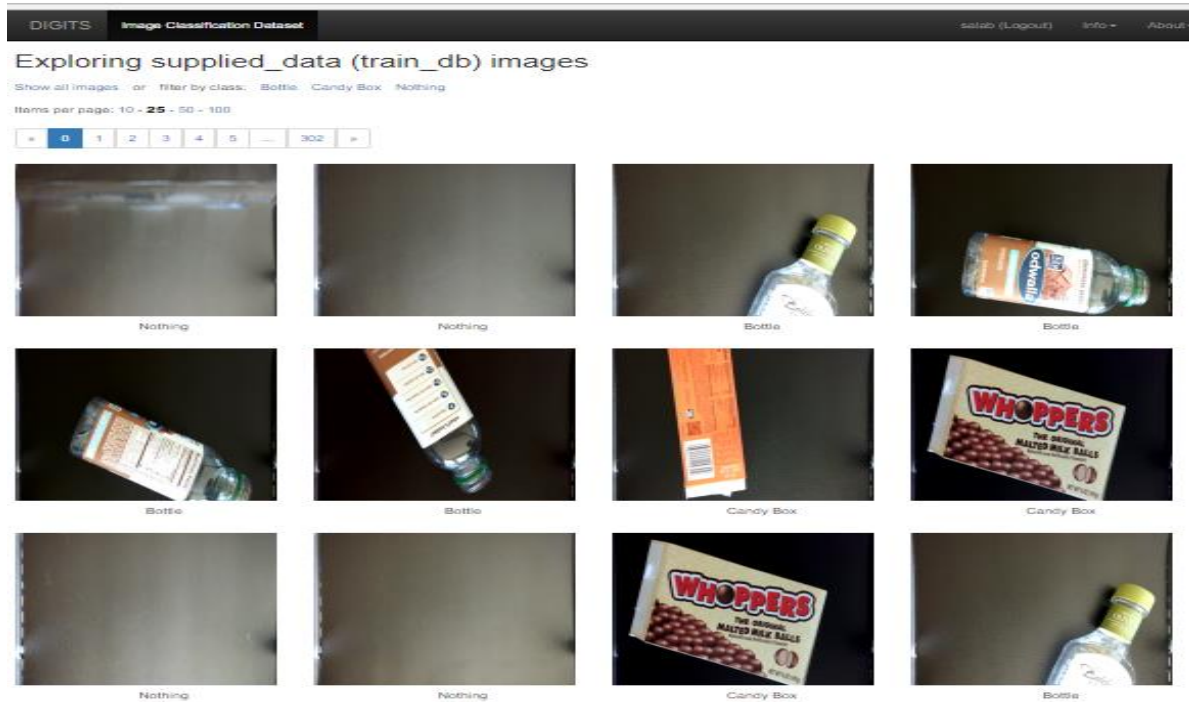


Figure 4: (P1) Supplied data sample images

4.0 RESULTS

The network with the supplied data was trained with GoogleNet and 100% accuracy achieved during the training. Upon evaluation, the model meets the required accuracy with 75.41% accuracy and the inference time was 5.06 ms which also meets the required inference time of below 10 ms.

On the other hand, the banknote's model was able to achieve accuracy of 99.87%.

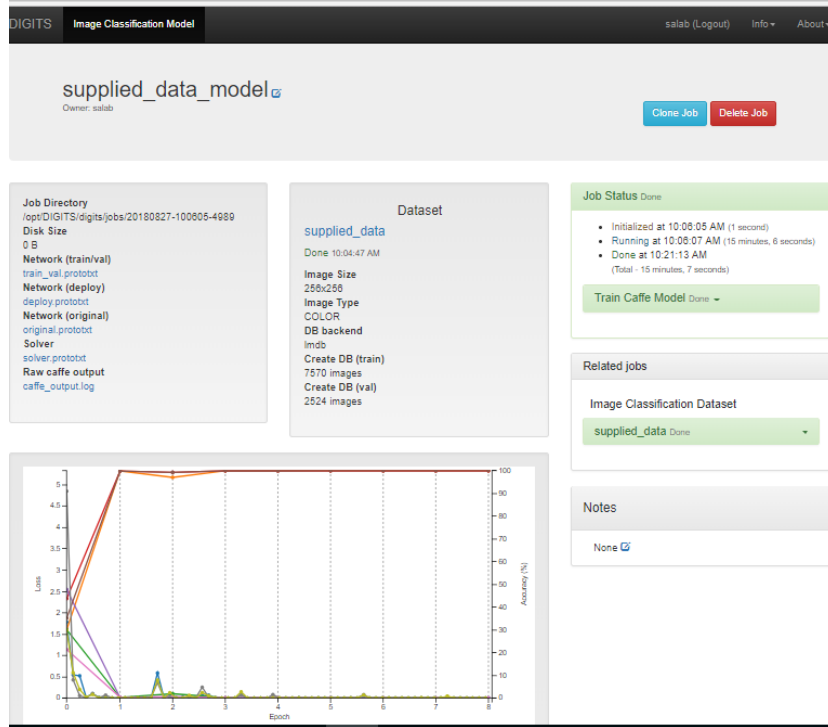


Figure 5: Suppled data training graph

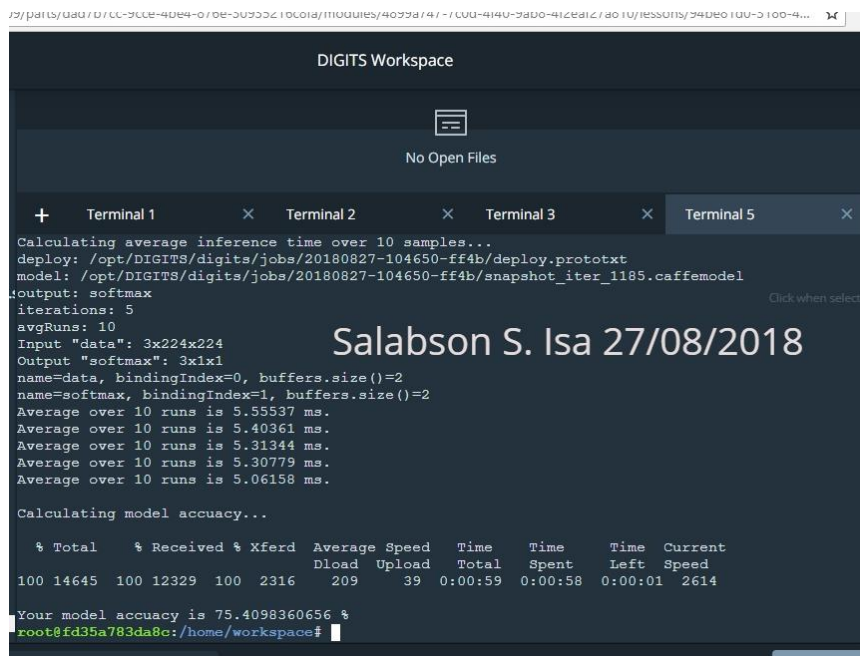


Figure 6: Supplied data evaluation accuracy

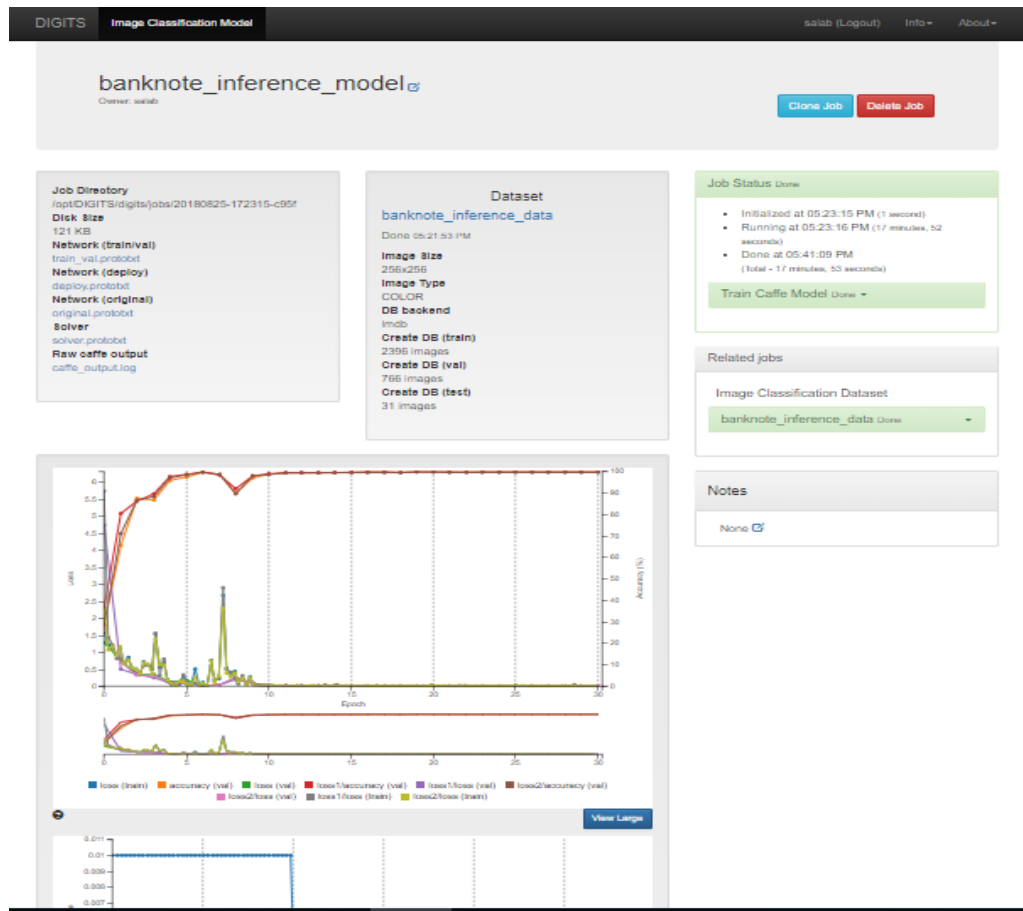


Figure 7: Banknote training graph

5.0 DISCUSSION

Both models achieved reasonable level of accuracy because considerable amount of sample image datasets were provided for the training and also right parameters were chosen. For example, initially when learning rate of 0.01 was used on the supplied data model, the model achieved accuracy of only 45.25% compared to 100% accuracy when learning rate of 0.001 was used. In the same vein, when Adam optimizer was used, upon evaluation only 73.77% accuracy was realized compared to 75.41% accuracy when Adam optimizer was replaced by Stochastic Gradient Descent.



Figure 8 When 0.01 learning rate was used

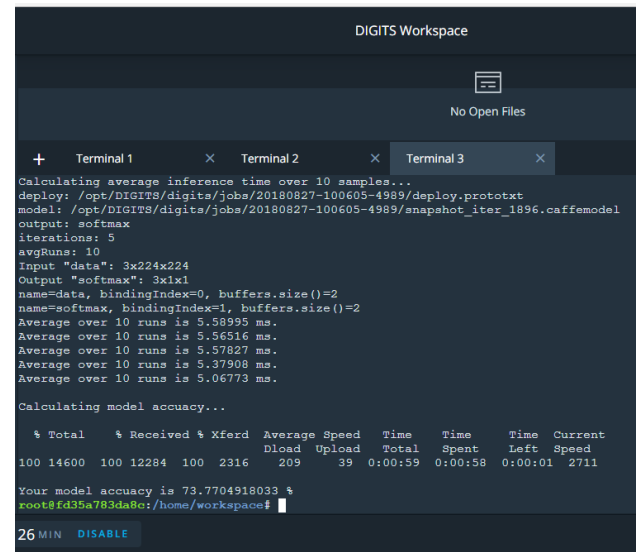


Figure 9: When Adam optimizer was used

6.0 CONCLUSION/FUTURE WORK

GoogleNet model was used to design two inference networks, one with P1 dataset provided by udacity and the required accuracy of at least 75% accuracy and inference time of below 10 ms are achieved. The other was built with own collected dataset and achieved accuracy of 99.887%.

Because the data collection was tedious, second model was limited to classifying only three banknotes from the available banknotes.

In future, all available banknotes will be included and the network will extrapolated to build counterfeit banknote detection system.

REFERENCES

- [1] Siddhart Das: CNN Architectures:LeNet, Alexnet, VGG, GoogleNet, ResNet and more.., [https://medium.com/@sideral/cnn-architectures-lenet-alexnet-vgg-googlenet-resnet-and more](https://medium.com/@sideral/cnn-architectures-lenet-alexnet-vgg-googlenet-resnet-and-more) , 2017.
- [2] Marcus Bloise: Data Aumentation, 2017.
- [3] A. Krizhevsky, I. Sutskever, and G. Hinton: ImageNet Classification with Deep Convolutional Networks, NIPS 2012