

Final Project - Individual report

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1. Project Introduction and outline of shared work.

The IAM Handwriting Dataset chosen for this project is comprised of a collection of handwritten passages by various writers. The goal of this project is to use deep learning to build classification model for handwritten recognition. There are totally 4899 image files in the dataset, written by 50 writers.

We divided this project into 4 steps: data pre-processing, network architecture design, training the network, evaluation of model performance.

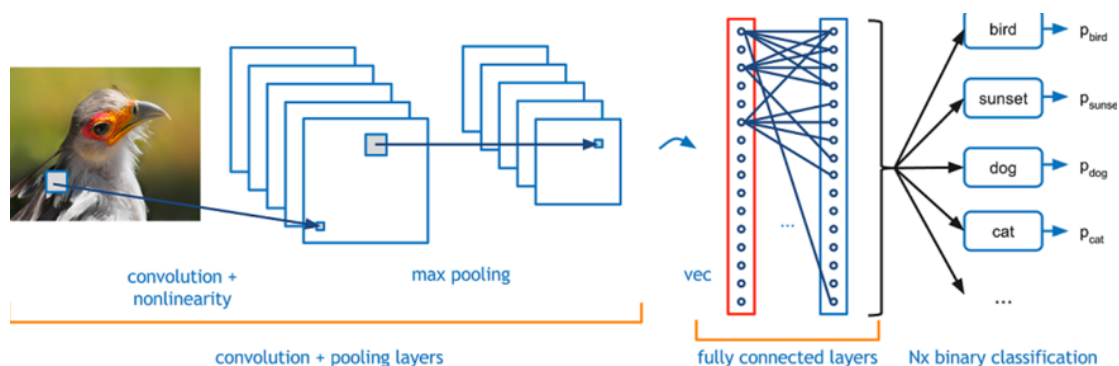
The code and the network architecture we worked on/designed together is the groupwork.py file, which is a 4 convolutional layer with SGD optimizer. We also met several times together to work on the final report, group presentation power-point. I gave details about my individual work in the following part.

2. Provide some background information on the development of the algorithm and include necessary equations and figures.

The reason why I chose CNN for this project is because it is most commonly applied to analyzing visual imagery, and our goal of this project is to classify writers based on their writing styles.

A CNN consists of an input and output layer, as well as multiple hidden layers. It reads the pixel value of images as input and goes through other layers to extract key features from the item in the image, then gives an output after classifying. Usually, the hidden layers of a CNN typically consist of convolutional layers, pooling layers and fully connected layers. These are what I leaned from the ML 2 class, and they (knowledge) were applied when I design the 4 convolutional layers model, which is discussed in the third question.

Here it is an example of how CNN works:



3. Describe the portion of the work that you did on the project in detail.

We worked on the 4-convolution-layer-SGD one together. The main work I did individually by myself is data pre-processing and 4 conv-layer network design.

- **Data Pre-Processing:**

I looked up the dataset and figured out what are these dashes connected name of the image files' meaning, and how many writers in the datasets in total. We decided to use only 50 writer's handwritten image to do classify in our projects.

I made change to the image files name and created a target list that has the writer number with respect to each image file in the dataset.

I also investigated on the image files, and found out those image sizes are quite different, so I decided to breakdown its image size to a standard one. There is also a reference chunk of code I found online for resizing and cropping smaller part of image. It's a generator function that was used to scan through each sentence and generate random patches with same patch size. We re-wrote this function to satisfy our model needs to process the image data. The final size of the cropped image is 113*113. Then, the images were read into pixel values. Last step was to normalize those pixel values.

One hot encoding was performed on target y as well.

- **Design of 4 layers of convolution neural network**

I designed the 4 layers of convolution neural network (Adam). Usually, the more number of convolutional layers, the better the information the model can capture, which leads to better prediction accuracy. The activation function I used is ReLU. The Maxpooling size I chose is 2*2 in all layers. The number of filters in each layer varies. In the first layer, I set up smaller number of filters, but I increased the number of filters in the later on layers so that the model can capture more detail information from the image. I also added dropout layer in the model, the dropout ratio is commonly set up to be 0.2-0.5, so that it can prevent the model to be overfitting.

The original plan in the training part was to change the number of samples in each epoch, the number of epoch and dropout rate so that I could compare the model performance.

4. Results. Describe the results of your experiments, using figures and tables wherever possible. Include all results (including all figures and tables) in the main body of the report, not in appendices. Provide an explanation of each figure and table that you include. Your discussions in this section will be most important part of the report.

Here is the summary of the model I designed:

Layer (type)	Output Shape	Param #
zero_padding2d_2 (ZeroPaddin	(None, 115, 115, 1)	0
lambda_2 (Lambda)	(None, 56, 56, 1)	0
conv1 (Conv2D)	(None, 28, 28, 32)	320
activation_8 (Activation)	(None, 28, 28, 32)	0
pool1 (MaxPooling2D)	(None, 14, 14, 32)	0
conv2 (Conv2D)	(None, 14, 14, 32)	9248
activation_9 (Activation)	(None, 14, 14, 32)	0
pool2 (MaxPooling2D)	(None, 7, 7, 32)	0
conv3 (Conv2D)	(None, 7, 7, 64)	18496
activation_10 (Activation)	(None, 7, 7, 64)	0
pool3 (MaxPooling2D)	(None, 3, 3, 64)	0
conv4 (Conv2D)	(None, 3, 3, 128)	73856
activation_11 (Activation)	(None, 3, 3, 128)	0
pool4 (MaxPooling2D)	(None, 1, 1, 128)	0
flatten_2 (Flatten)	(None, 128)	0
dropout_4 (Dropout)	(None, 128)	0
dense1 (Dense)	(None, 512)	66048
activation_12 (Activation)	(None, 512)	0
dropout_5 (Dropout)	(None, 512)	0
dense2 (Dense)	(None, 256)	131328
activation_13 (Activation)	(None, 256)	0
dropout_6 (Dropout)	(None, 256)	0
output (Dense)	(None, 50)	12850
activation_14 (Activation)	(None, 50)	0
Total params: 312,146		
Trainable params: 312,146		
Non-trainable params: 0		

This is a fully connected model. I added dropout layer into the model to prevent overfitting.

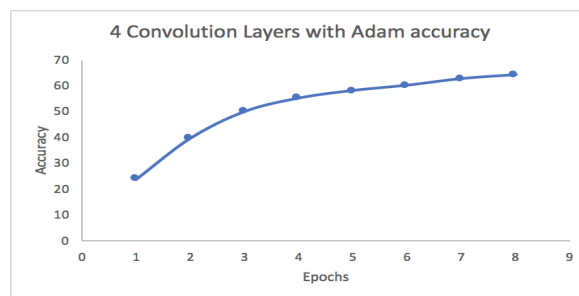
The running time of the 4-layer-network took longer time than 3 layers in average. The number of epochs I chose were 5 and 8. The number of samples to be used in each epoch that I chose were 500 and 1000. Here is the result I got when I set the num of epochs=8 and num of sample=1000.

It shows that the final accuracy is around 65%, which is not so good. The loss did not change a lot after the second epochs, and the accuracy was approaching stable after the sixth of epochs.

```
Epoch 1/8
1000/1000 [=====] - 2079s 2s/step - loss: 3.0565 - acc: 0.2387 - val_loss: 2.4752 - val_acc: 0.2715
Epoch 00001: saving model to check-01-2.4752.hdf5
Epoch 2/8
1000/1000 [=====] - 1885s 2s/step - loss: 2.0169 - acc: 0.3965 - val_loss: 1.6918 - val_acc: 0.4806
Epoch 00002: saving model to check-02-1.6918.hdf5
Epoch 3/8
1000/1000 [=====] - 1885s 2s/step - loss: 1.6261 - acc: 0.4993 - val_loss: 1.5528 - val_acc: 0.5243
Epoch 00003: saving model to check-03-1.5528.hdf5
Epoch 4/8
1000/1000 [=====] - 1888s 2s/step - loss: 1.4435 - acc: 0.5510 - val_loss: 1.3618 - val_acc: 0.5768
Epoch 00004: saving model to check-04-1.3618.hdf5
Epoch 5/8
1000/1000 [=====] - 1885s 2s/step - loss: 1.3455 - acc: 0.5802 - val_loss: 1.4098 - val_acc: 0.5608
Epoch 00005: saving model to check-05-1.4098.hdf5
Epoch 6/8
1000/1000 [=====] - 1885s 2s/step - loss: 1.2668 - acc: 0.6050 - val_loss: 1.3654 - val_acc: 0.5888
Epoch 00006: saving model to check-06-1.3654.hdf5
Epoch 7/8
1000/1000 [=====] - 1888s 2s/step - loss: 1.2024 - acc: 0.6257 - val_loss: 1.3765 - val_acc: 0.6301
Epoch 00007: saving model to check-07-1.3765.hdf5
Epoch 8/8
1000/1000 [=====] - 1876s 2s/step - loss: 1.1522 - acc: 0.6409 - val_loss: 1.0508 - val_acc: 0.6802
Epoch 00008: saving model to check-08-1.0508.hdf5
```

num of epochs	accuracy	loss
1	0.2387	3.0565
2	0.3965	2.0169
3	0.4993	1.6261
4	0.5510	1.4435
5	0.5802	1.3445
6	0.6050	1.2668
7	0.6257	1.2024
8	0.6409	1.1522

Here is the plot for the accuracy:



5. Summary and conclusions. Summarize the results you obtained, explain what have you learned, and suggest improvements that could be made in the future.

Compared with the neural network we designed (4-con-layer-SGD) together at the beginning, the performance of this (4-conv-layer-Adam) is much better. However, the accuracy is still not so high compared with the one with 3 conv layer Adam one.

From this process, I learned that when design the network for the deep learning model, it's essential to choose the right number of neurons, layers at the first few layers. In other hidden layers, dropout ratio is important, we should avoid to making it not too big. In the future, larger number of epochs and number of samples should be chosen in doing the modeling might help us to get better prediction result. Also, other optimizer should be tried in order to get the best accuracy too.

6. Calculate the percentage of the code that you found or copied from the internet.

The total number lines of code that I used as reference from internet is 31 lines. The number of lines I changed for this reference code part is 0. But the number of lines of code that I wrote in total by myself is 159, excluding comments lines and blank lines in the file. Hence the percentage is $31+15/(144+31)= 25.4\%$

7. References

Dataset:

<http://www.fki.inf.unibe.ch/databases/iam-handwriting-database>

Dropout ratio:

<https://machinelearningmastery.com/dropout-regularization-deep-learning-models-keras/>

Convolutional Networks:

<http://cs231n.github.io/convolutional-networks/>