**CMPT 412**

**Project 1**

**Digit recognition with convolutional neural networks**

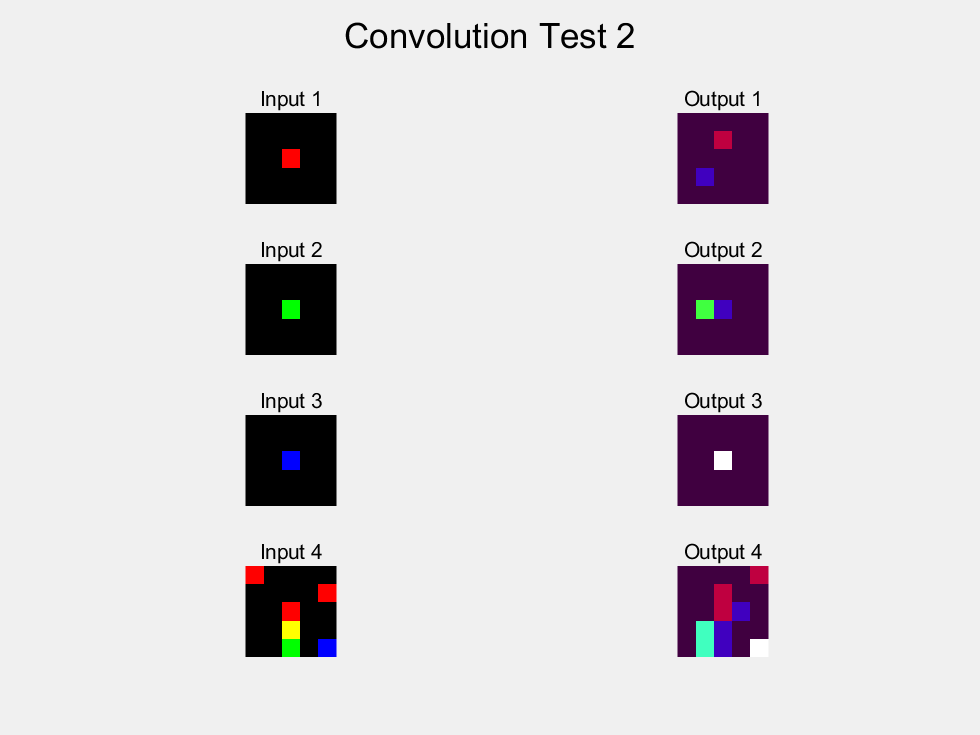
Instructor : Yasutaka Furukawa

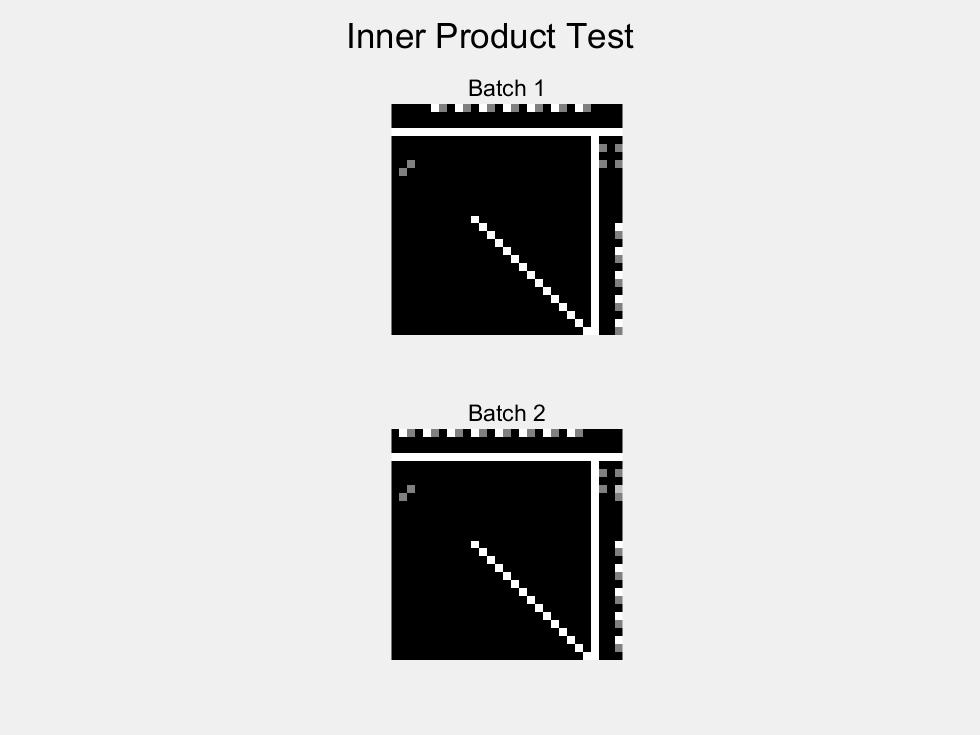
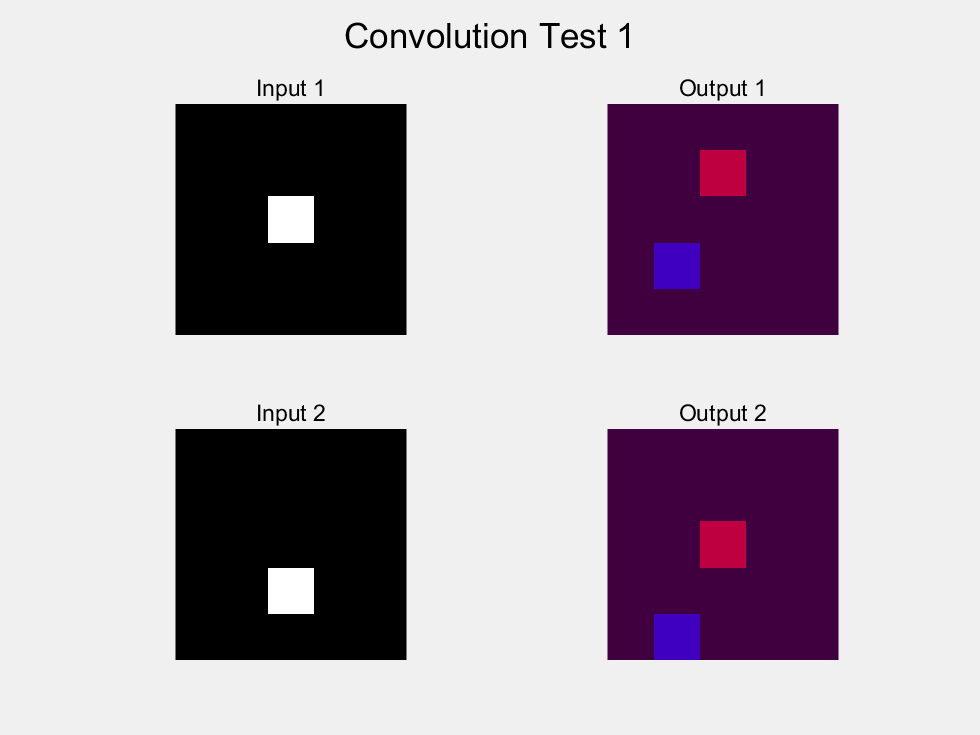
Name: Kaikun Fang

Student ID: 301416542

**Note:** The code inspiration and principles are derived from the videos and materials provided by the teacher. Discussions were also held with classmates(Cheng Hu 301435966).

**电脑萤幕画面

低可信度描述已自动生成Part 1: Forward Pass**

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**Part 2: Back propagation**

There is no visualization of the results in this section of the assignment requirements, the code has been completed and is in the corresponding file.

**Part 3: Training**

**Q3.1**

After about an hour of training, the following results were obtained. End at the test accuracy: 0.97 :

>> train\_lenet

cost = 0.273491 training\_percent = 0.910000

cost = 0.279565 training\_percent = 0.910000

cost = 0.176619 training\_percent = 0.920000

cost = 0.127344 training\_percent = 0.950000

cost = 0.191895 training\_percent = 0.960000

test accuracy: 0.944000

**… …**

cost = 0.069977 training\_percent = 1.000000

cost = 0.068312 training\_percent = 0.980000

cost = 0.063643 training\_percent = 0.980000

cost = 0.084625 training\_percent = 0.960000

cost = 0.083214 training\_percent = 0.980000

test accuracy: 0.970000

cost = 0.083081 training\_percent = 0.970000

cost = 0.026531 training\_percent = 1.000000

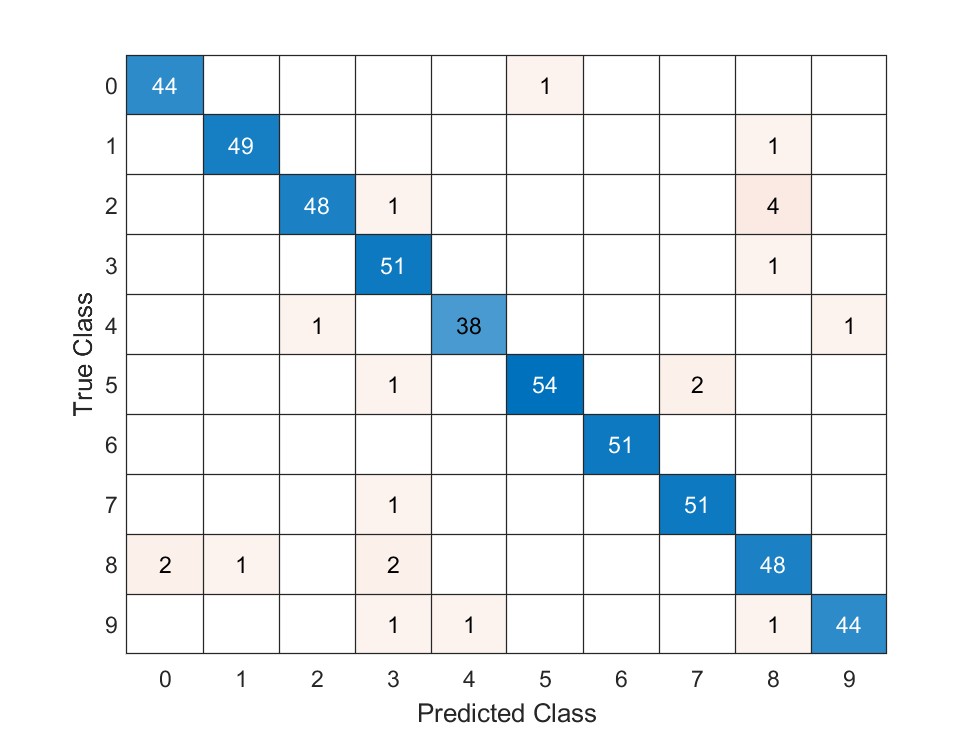
cost = 0.044653 training\_percent = 0.980000

cost = 0.056298 training\_percent = 0.980000

cost = 0.049833 training\_percent = 0.990000

test accuracy: 0.970000

**Q3.2**



The table shows that the two numbers most likely to be misidentified are ‘2’ and ‘8’.

‘2’ is easily recognized as ‘8’, which should be because ‘2’ may be too rounded in handwriting and cause the characteristics to match ‘8’.

‘8’ is easily recognized as ‘0’ and ‘3’ because ‘8’ itself is very similar to ‘3’. is very similar to ‘3’. Sometimes a handwritten ‘8’ may have a very small circle on the top or bottom, and after processing it may only have one circle left, so it will look very much like a ‘0’.

**Q3.3**

I wrote a script ('test\_Q3\_3.m') to automatically read and test my pre-segmented ten numbers, and the results are automatically saved in the '/results/test-3.3' folder. If you run the code repeatedly please delete the results of the last run first.

Origin:          

Output:          

Prediction: 9 1 2 3 4 5 6 7 8 9

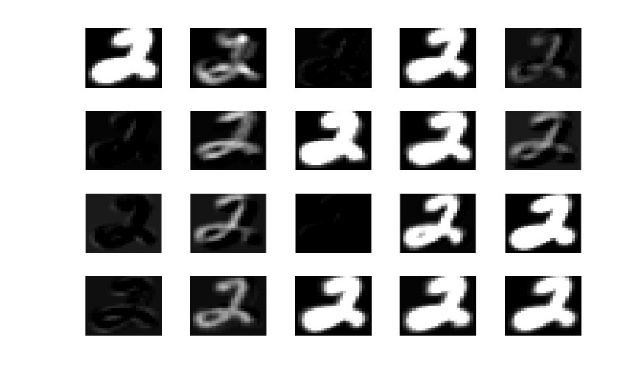
Probably because I subjectively selected some simple and clear images and pre-processed them for the test, so the recognition rate from the results is relatively high at ninety percent.

手机屏幕截图

中度可信度描述已自动生成**Part4:**



Figure 1 Origin Figure 2 Different

图片包含 游戏机

描述已自动生成

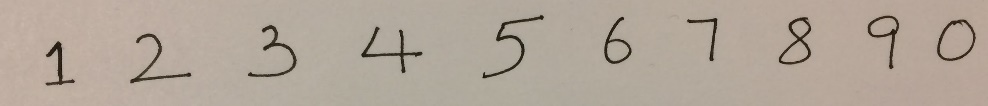
Figure 3CONV Figure 4 ReLU

I saved the generated images in the script into the '/result' folder. By observation, the results produced by both layers seem to be the same. Through research and search, it was found that this is because matlab's imshow turns negative numbers in CONV into 0 to display. And ReLU does exactly the same thing by turning the negative numbers in CONV into 0, so it causes the two plots to look the same. Since I created a new image(Figure 1) by extracting the negative numbers from CONV and converting the negative numbers in it to 1 to show it, you can see which data in CONV is modified by ReLU.

**Part 5: Image Classification**

The script I wrote (ec.m) automatically reads and identifies the four images provided by the teacher and outputs the results to the corresponding "/results/image#" folder. When running the code repeatedly, please delete the results of the last run first. The results of my run are as follows:

**Image1:**

Origin: 

Output:          

Prediction: 1 2 3 4 7 5 3 8 7 0

Correct Rate: 6/10

**Image2:**

Origin: 文本, 信件

描述已自动生成

Output:          

Prediction: 1 2 3 9 5 5 3 3 7 0

Correct Rate: 5/10

**Image3:**

Origin: 卡通人物

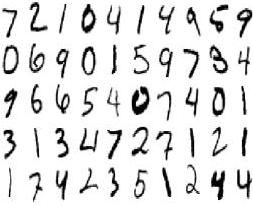
低可信度描述已自动生成

Output: 

Prediction: 6 0 6 3 6

Correct Rate: 3/5

**Image4:**

Origin: 

Output:           

Prediction:7 2 1 9 4 1 4 9 X 7

0 6 4 9 1 5 9 7 5 4

7 6 6 5 4 9 7 4 0 1

7 1 3 4 7 2 7 1 2 1

1 7 4 2 1 5 1 2 4 4

Correct Rate: 40/50

**Summary:**

The overall results showed that 54 out of 75 numbers were correctly identified, with an accuracy rate of about 72%. Interestingly, in image4, the number "5" in the top right of the original image was not connected together when it was written, so it was cut into two parts and recognized separately, which led to the wrong result. Secondly, the recognition rate of 0, 3 and 9 is not very accurate.