SC 4001 CE/CZ 4042 Neural Network and Deep Learning

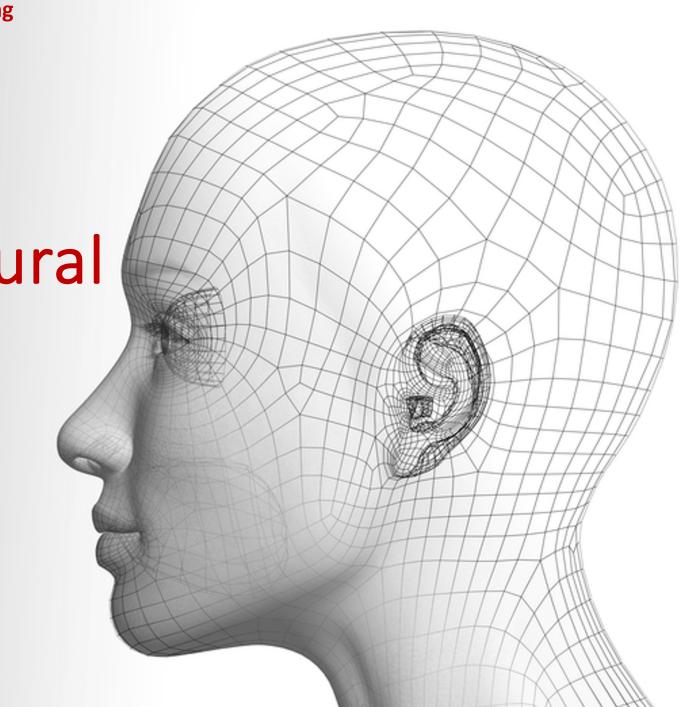
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Tutorial 6
Convolutional Neural
Networks I

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1. The first hidden layer of a convolution neural network (CNN) has a convolution layer consisting of two feature maps with filters w_1 and w_2 and biases = 0.1, and neurons having sigmoid activation functions, and a pooling layer with a pooling windows of size 2x2:

$$\mathbf{w}_1 = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \text{ and } \mathbf{w}_2 = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix}.$$

The input layer is of 6x6 size and receives an input image I:

$$I = \begin{pmatrix} 0.7 & 0.1 & 0.2 & 0.3 & 0.3 & 05 \\ 0.8 & 0.1 & 0.3 & 0.5 & 0.1 & 0.0 \\ 1.0 & 0.2 & 0.0 & 0.3 & 0.2 & 0.7 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \\ 1.0 & 0.1 & 0.4 & 0.5 & 0.2 & 0.8 \end{pmatrix}$$

- a. Find the outputs at the first convolution layer if
 - i. padding = 0 and strides = (1,1)
 - ii. padding = 1 and strides = (2,2)
- b. Find the outputs at the first pooling layer for Part (a), assuming strides of (2, 2) and pooling is
 - i. max pooling
 - ii. mean pooling

Filters=weights

t6q1.ipynb

$$\mathbf{W}_{1} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}, \mathbf{W}_{2} = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \text{ and } b = 0.1.$$

$$\mathbf{I} = \begin{pmatrix} 0.7 & 0.1 & 0.2 & 0.3 & 0.3 & 05 \\ 0.8 & 0.1 & 0.3 & 0.5 & 0.1 & 0.0 \\ 1.0 & 0.2 & 0.0 & 0.3 & 0.2 & 0.7 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \end{pmatrix}$$

Synaptic inputs to the feature map with filter w_1 :

$$\boldsymbol{u}_1 = Conv(\boldsymbol{I}, \boldsymbol{w}_1) + b_1$$

Padding = 0 and strides = (1,1)

$$u_1(1,1) = 0.7 \times 0 + 0.1 \times 1 + 0.2 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.3 \times 1 + 1.0 \times 1 + 0.2 \times 1 + 0.0 \times 0 + 0.1 = 2.7$$

 $u_1(1,2) = 0.1 \times 0 + 0.2 \times 1 + 0.3 \times 1 + 0.1 \times 1 + 0.3 \times 0 + 0.5 \times 1 + 0.2 \times 1 + 0.0 \times 1 + 0.3 \times 0 + 0.1 = 1.4$
 $u_1(2,1) = 0.8 \times 0 + 0.1 \times 1 + 0.3 \times 1 + 1.0 \times 1 + 0.2 \times 0 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 1 + 0.5 \times 0 + 0.1 = 2.4$

$$\boldsymbol{u}_1 = \begin{pmatrix} 2.7 & 1.4 & 1.4 & 1.9 \\ 2.4 & 2.0 & 2.0 & 2.1 \\ 1.7 & 2.0 & 2.6 & 2.6 \\ 2.8 & 2.0 & 3.1 & 2.0 \end{pmatrix}$$

Similarly,
$$\mathbf{u}_2 = Conv(\mathbf{I}, \mathbf{w}_2) + b_2 = \begin{pmatrix} 0.3 & 0.0 & -0.2 & 2.0 \\ 0.3 & 0.4 & 0.6 & 0.8 \\ -0.1 & 0.8 & 1.1 & -0.3 \\ 0.2 & -0.1 & -0.2 & 0.3 \end{pmatrix}$$

Feature maps at the convolutional layer

$$\mathbf{y}_1 = f(\mathbf{u}_1) = \frac{1}{1 + e^{-\mathbf{u}_1}} = \begin{pmatrix} 0.94 & 0.8 & 0.8 & 0.87 \\ 0.92 & 0.88 & 0.88 & 0.89 \\ 0.85 & 0.88 & 0.93 & 0.93 \\ 0.94 & 0.88 & 0.96 & 0.88 \end{pmatrix}$$

$$\mathbf{y}_2 = f(\mathbf{u}_2) = \frac{1}{1 + e^{-\mathbf{u}_2}} = \begin{pmatrix} 0.57 & 0.5 & 0.45 & 0.55 \\ 0.57 & 0.60 & 0.65 & 0.69 \\ 0.48 & 0.69 & 0.75 & 0.43 \\ 0.55 & 0.48 & 0.45 & 0.57 \end{pmatrix}$$

Output spatial size = (Input size – Filter size +2(Padding))/Stride + 1 = (6-3)/1 + 1 = 4

Feature maps at the convolutional layer:

$$\mathbf{y} = \begin{pmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{pmatrix} = \begin{pmatrix} 0.94 & 0.8 & 0.8 & 0.87 \\ 0.92 & 0.88 & 0.88 & 0.89 \\ 0.85 & 0.88 & 0.93 & 0.93 \\ 0.94 & 0.88 & 0.96 & 0.88 \\ 0.57 & 0.5 & 0.45 & 0.55 \\ 0.57 & 0.60 & 0.65 & 0.69 \\ 0.48 & 0.69 & 0.75 & 0.43 \\ 0.55 & 0.48 & 0.45 & 0.57 \end{pmatrix}$$

Pooling 2×2 and strides = 2

Max-pooling:

$$o = \begin{pmatrix} \begin{pmatrix} 0.94 & 0.89 \\ 0.94 & 0.96 \end{pmatrix} \\ \begin{pmatrix} 0.60 & 0.69 \\ 0.69 & 0.75 \end{pmatrix} \end{pmatrix}$$

Mean-pooling:

$$p_{ave} = \begin{pmatrix} 0.88 & 0.86 \\ 0.89 & 0.92 \end{pmatrix} \\ \begin{pmatrix} 0.56 & 0.58 \\ 0.55 & 0.55 \end{pmatrix}$$

$$\mathbf{w}_{1} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}, \mathbf{w}_{2} = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \text{ and } b = 0.1.$$

$$\mathbf{I} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0.7 & 0.1 \\ 0.8 & 0.1 & 0.2 & 0.3 & 0.3 & 0.5 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \\ 1.0 & 0.1 & 0.4 & 0.5 & 0.2 & 0.8 \end{pmatrix}$$

Synaptic inputs to the feature map with filter w_1 :

$$\boldsymbol{u}_1 = Conv(\boldsymbol{I}, \boldsymbol{w}_1) + b_1$$

Padding = 1 and strides = (2,2)

$$u_1(1,1) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.0 \times 1 + 0.0 \times 1 + 0.7 \times 0 + 0.1 \times 1 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.1 = 1.0$$

$$\boldsymbol{W}_{1} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}, \boldsymbol{W}_{2} = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \text{ and } b = 0.1.$$

$$\boldsymbol{I} = \begin{pmatrix} 0.7 & 0.1 & 0.2 & 0.3 & 0.3 & 05 \\ 0.8 & 0.1 & 0.2 & 0.0 & 0.3 & 0.2 & 0.7 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \\ 1.0 & 0.1 & 0.4 & 0.5 & 0.2 & 0.8 \end{pmatrix}$$

Synaptic inputs to the feature map with filter w_1 :

$$\boldsymbol{u}_1 = Conv(\boldsymbol{I}, \boldsymbol{w}_1) + b_1$$

Padding =
$$1$$
 and strides = $(2,2)$

$$\boldsymbol{u}_{1}(1,1) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.0 \times 1 + 0.7 \times 0 + 0.1 \times 1 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.1 = 1.0$$

$$u_1(1,2) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.1 \times 1 + 0.2 \times 0 + 0.3 \times 1 + 0.1 \times 1 + 0.3 \times 1 + 0.5 \times 0 + 0.1 = 0.9$$

$$\mathbf{W}_{1} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}, \mathbf{W}_{2} = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \text{ and } b = 0.1.$$

$$\mathbf{I} = \begin{pmatrix} 0.7 & 0.1 & 0.2 & 0.3 & 0.3 & 0.5 \\ 0.8 & 0.1 & 0.3 & 0.5 & 0.1 & 0.0 \\ 1.0 & 0.2 & 0.0 & 0.3 & 0.2 & 0.7 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \\ 1.0 & 0.1 & 0.4 & 0.5 & 0.2 & 0.8 \end{pmatrix}$$

Synaptic inputs to the feature map with filter w_1 :

$$\boldsymbol{u}_1 = Conv(\boldsymbol{I}, \boldsymbol{w}_1) + b_1$$

Padding = 1 and strides = (2,2)

$$u_1(1,1) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.0 \times 1 + 0.7 \times 0 + 0.1 \times 1 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.1 = 1.0$$

 $u_1(1,2) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.1 \times 1 + 0.2 \times 0 + 0.3 \times 1 + 0.1 \times 1 + 0.3 \times 1 + 0.5 \times 0 + 0.1 = 0.9$
 $u_1(1,3) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.3 \times 1 + 0.3 \times 0 + 0.5 \times 1 + 0.5 \times 1 + 0.1 \times 1 + 0.0 \times 0 + 0.1 = 1.5$

$$\mathbf{w}_1 = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}, \mathbf{w}_2 = \begin{pmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix}$$
 and $b = 0.1$.

$$I = \begin{bmatrix} 0.7 & 0.1 & 0.2 & 0.3 & 0.3 & 05 \\ 0.8 & 0.1 & 0.3 & 0.5 & 0.1 & 0.0 \\ 1.0 & 0.2 & 0.0 & 0.3 & 0.2 & 0.7 \\ 0.8 & 0.1 & 0.5 & 0.6 & 0.3 & 0.4 \\ 0.1 & 0.0 & 0.9 & 0.3 & 0.3 & 0.2 \\ 1.0 & 0.1 & 0.4 & 0.5 & 0.2 & 0.8 \end{bmatrix}$$

Synaptic inputs to the feature map with filter w_1 :

$$\boldsymbol{u}_1 = Conv(\boldsymbol{I}, \boldsymbol{w}_1) + b_1$$

Padding = 1 and strides = (2,2)

$$u_1(1,1) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.0 \times 1 + 0.7 \times 0 + 0.1 \times 1 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.1 = 1.0$$

 $u_1(1,2) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.1 \times 1 + 0.2 \times 0 + 0.3 \times 1 + 0.1 \times 1 + 0.3 \times 1 + 0.5 \times 0 + 0.1 = 0.9$
 $u_1(1,3) = 0.0 \times 0 + 0.0 \times 1 + 0.0 \times 1 + 0.3 \times 1 + 0.3 \times 0 + 0.5 \times 1 + 0.5 \times 1 + 0.1 \times 1 + 0.0 \times 0 + 0.1 = 1.5$
 $u_1(2,1) = 0.0 \times 0 + 0.8 \times 1 + 0.1 \times 1 + 0.0 \times 1 + 1.0 \times 0 + 0.2 \times 1 + 0.0 \times 1 + 0.8 \times 1 + 0.1 \times 0 + 0.1 = 2.0$

$$\boldsymbol{u}_1 = \begin{pmatrix} 1.0 & 0.9 & 1.5 \\ 2.0 & 2.0 & 2.1 \\ 2.0 & 2.0 & 2.0 \end{pmatrix}$$

Similarly,
$$\mathbf{u}_2 = Conv(\mathbf{I}, \mathbf{w}_2) + b_2 = \begin{pmatrix} 1.0 & 1.0 & 0.7 \\ 0.1 & 0.4 & 0.8 \\ 0.3 & -0.1 & 0.3 \end{pmatrix}$$

Feature maps at the convolutional layer

$$\mathbf{y}_1 = f(\mathbf{u}_1) = \frac{1}{1 + e^{-\mathbf{u}_1}} = \begin{pmatrix} 0.73 & 0.71 & 0.82 \\ 0.88 & 0.88 & 0.89 \\ 0.88 & 0.88 & 0.88 \end{pmatrix}$$

$$\mathbf{y}_2 = f(\mathbf{u}_2) = \frac{1}{1 + e^{-\mathbf{u}_2}} = \begin{pmatrix} 0.73 & 0.73 & 0.67 \\ 0.52 & 0.60 & 0.69 \\ 0.57 & 0.48 & 0.57 \end{pmatrix}$$

Feature maps at the convolutional layer:

$$\mathbf{y} = \begin{pmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{pmatrix} = \begin{pmatrix} 0.73 & 0.71 & 0.82 \\ 0.88 & 0.88 & 0.89 \\ 0.88 & 0.88 & 0.88 \\ 0.73 & 0.73 & 0.67 \\ 0.52 & 0.60 & 0.69 \\ 0.57 & 0.48 & 0.57 \end{pmatrix}$$

Pooling VALID 2x2 strides = 2:

Max-pooling:

$$o = \begin{pmatrix} (0.88) \\ (0.73) \end{pmatrix}$$

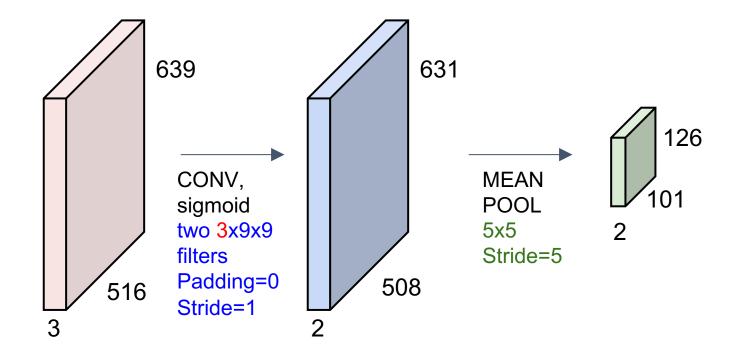
Mean-pooling:

$$p_{ave} = \begin{pmatrix} (0.80) \\ (0.65) \end{pmatrix}$$

- 2. Given '3wolfmoon.jpg' color image of size 639×516 .
 - a. Initialize weights and biases of a convolutional layer with two kernels of size 9×9 . Note that the input image is in color and has three channels.
 - b. Display the feature maps at the convolution layer, assuming sigmoid activation functions. Use VALID padding and strides = 1.
 - c. Display the outputs of a mean pooling layer with a pooling window size 5×5 and strides = 5.







```
# open image and normalize
img = Image.open('3wolfmoon.jpg')
img = np.asarray(img, dtype='float32') / 256.
print(img.shape)
FileNotFoundError
                                          Traceback (most recent call last)
<ipython-input-2-becdc6409bf5> in <module>()
      1 # open image and normalize
---> 2 img = Image.open('3wolfmoon.jpg')
      3 img = np.asarray(img, dtype='float32') / 256.
      4 print(img.shape)
/usr/local/lib/python3.7/dist-packages/PIL/Image.py in open(fp, mode)
   2841
   2842
            if filename:
                fp = builtins.open(filename, "rb")
-> 2843
                exclusive fp = True
   2844
   2845
FileNotFoundError: [Errno 2] No such file or directory: '3wolfmoon.jpg'
```

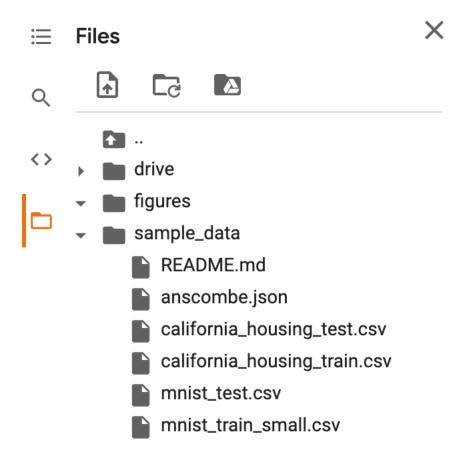
You need to mount Google Drive locally and use the correct path

https://colab.research.google.com/notebooks/io.ipynb#scrollTo=u22w3BFiOveA

```
# mount google drive
from google.colab import drive
drive.mount('/content/drive')

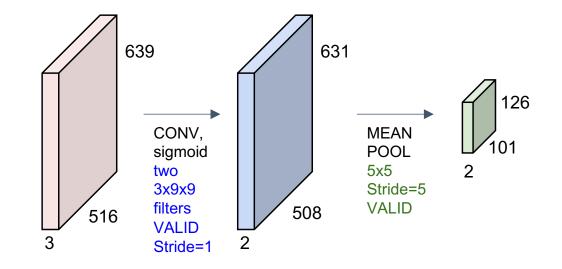
Mounted at /content/drive
```

```
# open image and normalize
img = Image.open('/content/drive/MyDrive/3wolfmoon.jpg')
img = np.asarray(img, dtype='float32') / 256.
print(img.shape)
(639, 516, 3)
```



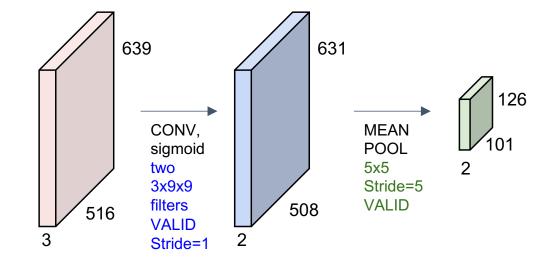
You can copy the path from the browser

Plot the three channels of the input image





Plot the original image and the first and second components of conv output





Original image

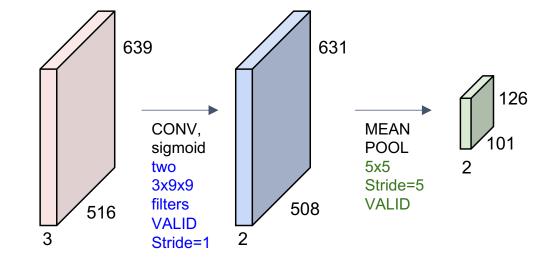


Feature map 1



Feature map 2

Plot the original image and the first and second components of pooling output





Original image



Pooled feature map 1



Pooled feature map 2

3. Design a CNN with one hidden layer to recognize digit images in MNIST database:

http://yann.lecun.com/exdb/mnist/

The convolution layer consists of 25 filters of dimensions 9x9 and the pooling layer has a pooling window size 4x4. Assume VALID padding and default strides for both convolution and pooling layer. Train the network with mini batch gradient decent learning with learning factor $\alpha = 10^{-3}$ and batch size = 128.

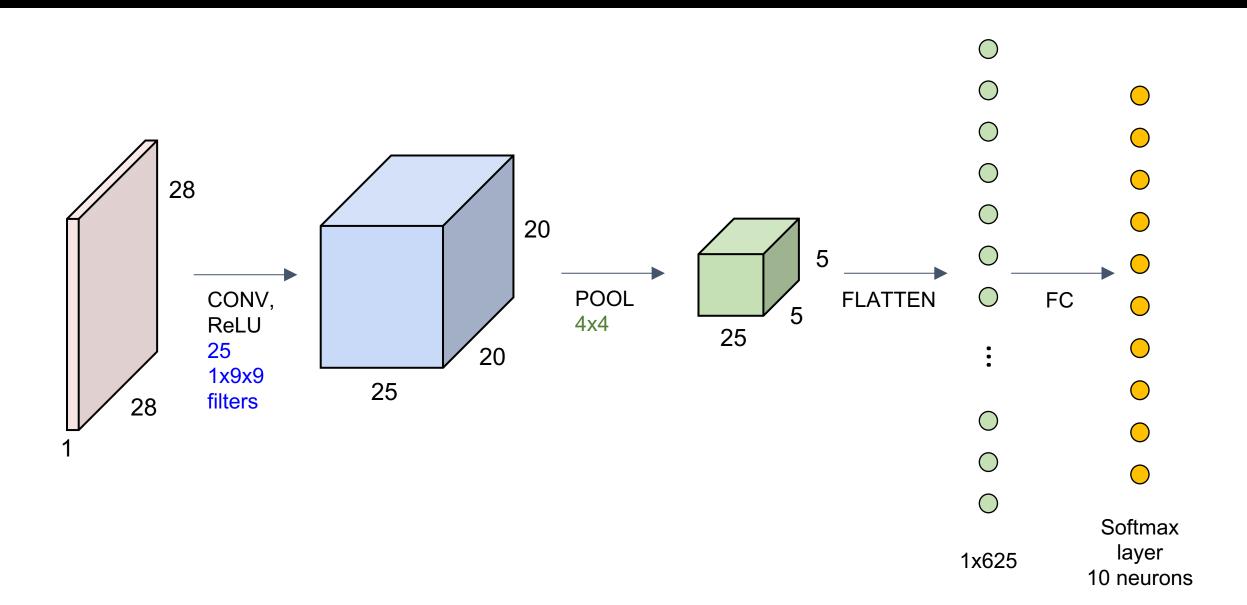
Plot

- a. The training and test errors against learning epochs.
- b. Final filter weights
- c. Feature maps at the convolution and pooling layers for a representative test pattern
- d. Repeat training by introducing decay parameter $\beta=10^{-6}$ and momentum term with $\gamma=0.5$, and compare the learning curves

t6q3a.ipynb

MNIST

- Size-normalized and centred 1x28x28 = 784 inputs
- Training set = 60,000 images
- Testing set = 10,000 images



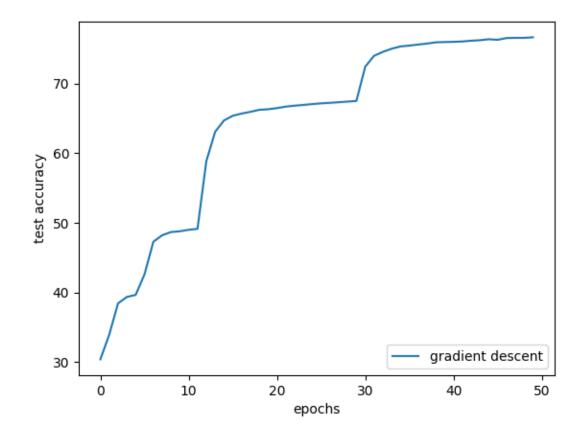
Change runtime type

Python 3 Hardware accelerator ? CPU T4 GPU A100 GPU V100 GPU TPU Want access to premium GPUs? Purchase additional compute units

Cancel Save

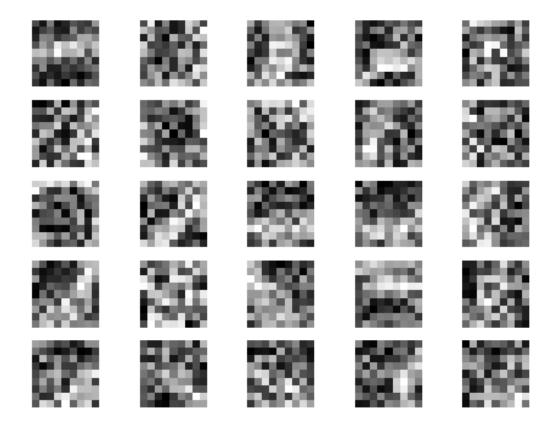
Question 3a

Plot the learning curve (I only show the test curve) - you should be able get better learning curve if you run for more epochs

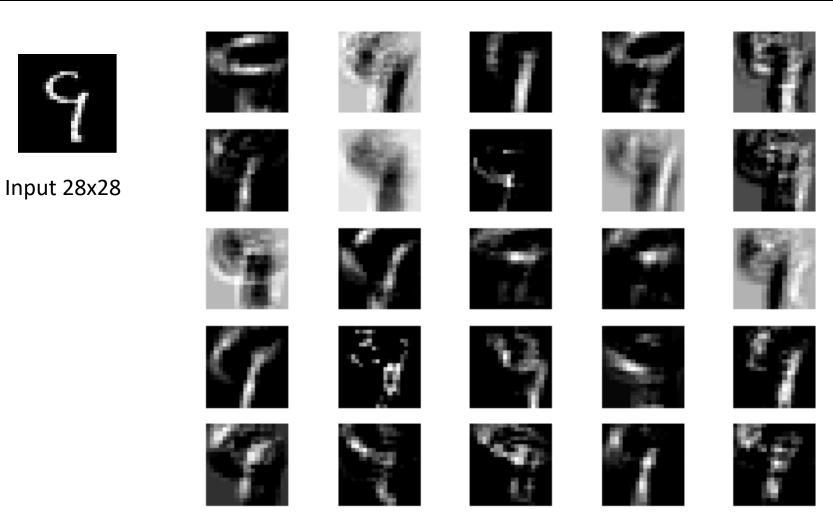


Question 3b

Plot filters learned in the conv layer



Question 3c



Feature maps of pooling layer 25x5x5

Feature maps of the convolution layer 25x20x20

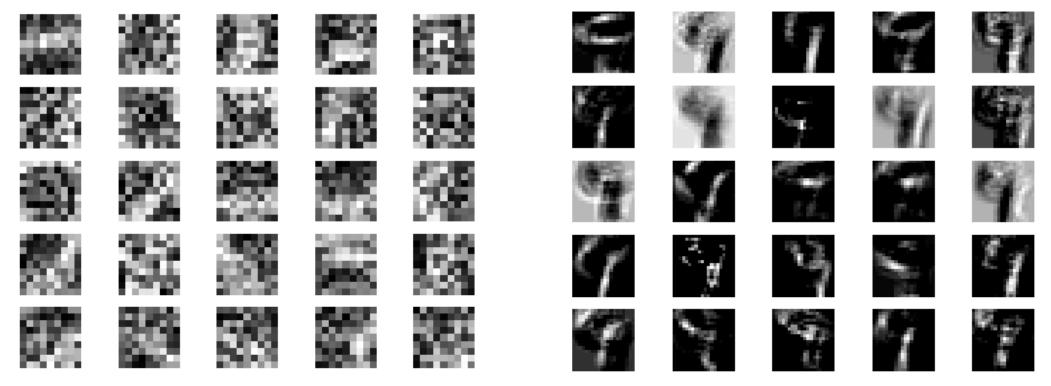
Question 3c



Compare the filters learned and the corresponding feature maps.

What do you see?

Input 28x28



Filters learned in the conv layer

Feature maps of the convolution layer 25x20x20

Question 3d

Gradient descent:

$$W = W - \alpha \nabla_W J$$

Weight decay (regularization):

$$J_1(\boldsymbol{W}, \boldsymbol{b}) = J(\boldsymbol{W}, \boldsymbol{b}) + \beta_2 \sum_{ij} (w_{ij})^2$$

Weight decay:

$$W = W - \alpha(\nabla_W J + \beta W)$$

Sgd with momentum:

$$V = \gamma V - \alpha \nabla_W J$$
$$W = W + V$$

Sgd with decay and momentum:

$$V = \gamma V - \alpha (\nabla_W J + \beta W)$$
$$W = W + V$$

