### Lab: CudaVision – Learning Vision Systems on Graphics Cards (MA-INF 4308)

Assignment 4

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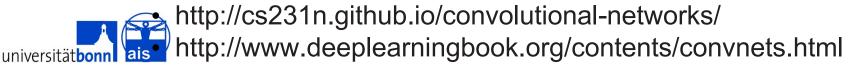


## Convolutional Neural Networks with MNIST dataset

- Goal: How to build a Convolutional Neural Network (CNN) Models with TensorFlow
- Let's learn basic components to build CNN layer

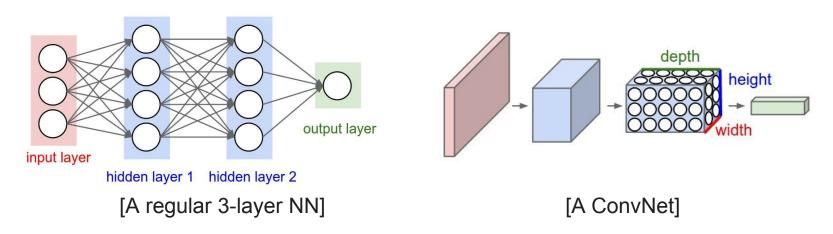
```
tf.reshape(), tf.nn.conv2d(), tf.nn.bias_add(),
   tf.nn.relu(), tf.nn.max pool()
```

- Design parameters
  - patch\_size, strides, padding, number of filters
- Your task is,
  - Build your CNN models
  - Find the best performance of MNIST classifier
  - Visualize trained filters
- Theoritical reference



#### **Convolutional Networks (ConvNet)**

- Full connectivity in regular neural networks don't scale well to full images.
- Unlike a regular NN, the layers of a ConvNet have neurons arranged in 3 dimensions: width, height, channel.



 Three main types of layers to build ConvNet: Convolutional Layer, Pooling Layer, and Fully-connected layer



#### **Example**

- Input raw rgb pixel values of the image: [32x32x3]
- CONV layer will compute the output of neurons that are connected to local regions in the input : [32x32x12], if use 12 filters.
- RELU layer will apply an elementwise activation fulction [32x32x12]
- POOL layer will perform a downsampling operation along the spatial dimensions: [16x16x12], if use window size 2
- FC (fully-connected) layer will compute the class scores [1x1x10]
- Learning parameters
  - Weights and biasses in CONV/FC layers
- Hyper parameters of a ConvLayer
  - The number of filters
  - Filter size
  - Stride size: the number of pixels to move the filters at a time
  - Zero-padding size



# 1 ConvLayer implementation for MNIST (24 x 24 x 1) data

in CNNmodel\_single\_MNIST.ipynb

```
n input width = 28
n input height = 28
n input channel = 1
n convl patch size = 3
n conv1 filter = 64
n output = 10 # e.g. MNIST total classes (0-9 digits)
# tf Graph input
x = tf.placeholder(tf.float32, [None, n input width * n input height])
y = tf.placeholder(tf.float32, [None, n output])
wcl = tf.Variable(tf.random normal([n conv1 patch size, n conv1 patch size, n input channel, n conv1 filter], stddev=0.1))
bc1 = tf.Variable(tf.random normal([n conv1 filter], stddev=0.1))
wf1 = tf.Variable(tf.random normal([(n input width/2)*(n input height/2)*n conv1 filter, n output], stddev=0.1))
bf1 = tf.Variable(tf.random normal([n output], stddev=0.1))
# Reshape input
input r = tf.reshape(x, shape=[-1, n input width, n input width, 1])
# Convolution
conv = tf.nn.conv2d(input r, wcl, strides=[1, 1, 1, 1], padding='SAME')
# Add-bias
bias = tf.nn.bias add(conv, bc1)
# Pass ReLu
relu = tf.nn.relu(bias)
# Max-pooling
pool = tf.nn.max pool(relu, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME')
# Vectorize
dense = tf.reshape(pool, [-1, wfl.get shape().as list()[0]])
# Fully-connected layer
out = tf.add(tf.matmul(dense, wf1), bf1)
cost = tf.reduce mean(tf.nn.softmax cross entropy with logits(out, y))
```



#### Data reshaping

- tf.reshape(tensor, shape, name=None)
  - Given tensor, this operation returns a tensor that has the same values as tensor with shape shape.
  - If one component of shape is the special value -1, the size of that dimension is computed so that the total size remains constant. In particular, a shape of [-1] flattens into 1-D. At most one component of shape can be -1.
  - If shape is 1-D or higher, then the operation returns a tensor with shape filled with the values of tensor. In this case, the number of elements implied by shape must be the same as the number of elements in tensor.



#### **Assignment 4**

- Finding your best CNN model for MNIST classification
  - Train the given single ConvNet with MNIST dataset
    - Load MNIST
    - Construct a optimizer using tf.train.AdamOptimizer
    - Train the model by changing learning\_rate, training\_epochs, batch size, the number and size of filters,
    - What is the test accuracy of the given model ? (I got 0.979)
  - Build and train deeper CNNs with tf.nn.dropout
    - Add more Conv layers
    - Add dropout function after each pooling layer
  - What is your best performance with a multi-layered CNN?

Visualize images of input image, output image of each Conv filter, weight

images of each Conv filter

