Applied Deep Learning with Tensorflow and PyTorch, Chapter 6, Week 2

Batch Normalization and LeNet

LeNet is among the first published CNNs to capture wide attention for its performance on computer vision tasks. The model was introduced by and named for Yann LeCun.

Note that the architecture below is LeNet architecture with additional batch normalization layers and the original LeNet model does not contain batch normalization layers. The original LeNet model is proposed in 1989, way before the batch normalization is proposed in 2015. We include the batch normalization layers to LeNet for illustrative purposes regarding how these layers can be used as part of a model.

LeNet with Batch Normalization in PyTorch

LeNet with Batch Normalization in Tensorflow

```
def net():
   return tf.keras.models.Sequential([
        tf.keras.layers.Conv2D(filters=6, kernel_size=5,
                               input_shape=(28, 28, 1)),
       tf.keras.layers.BatchNormalization(),
        tf.keras.layers.Activation('sigmoid'),
       tf.keras.layers.MaxPool2D(pool_size=2, strides=2),
       tf.keras.layers.Conv2D(filters=16, kernel_size=5),
        tf.keras.layers.BatchNormalization(),
       tf.keras.layers.Activation('sigmoid'),
        tf.keras.layers.MaxPool2D(pool_size=2, strides=2),
       tf.keras.layers.Flatten(),
       tf.keras.layers.Dense(120),
       tf.keras.layers.BatchNormalization(),
        tf.keras.layers.Activation('sigmoid'),
       tf.keras.layers.Dense(84),
        tf.keras.layers.BatchNormalization(),
        tf.keras.layers.Activation('sigmoid'),
        tf.keras.layers.Dense(10),])
```

AlexNet

AlexNet, which employed an 8-layer CNN, won the ImageNet Large Scale Visual Recognition Challenge 2012 by a phenomenally large margin. This network showed, for the first time, that the features obtained by learning can transcend manually-designed features, breaking the previous paradigm in computer vision.

AlexNet in PyTorch

```
net = nn.Sequential(
    nn.Conv2d(1, 96, kernel_size=11, stride=4, padding=1), nn.ReLU(),
    nn.MaxPool2d(kernel_size=3, stride=2),
    nn.Conv2d(96, 256, kernel_size=5, padding=2), nn.ReLU(),
    nn.MaxPool2d(kernel_size=3, stride=2),
    nn.Conv2d(256, 384, kernel_size=3, padding=1), nn.ReLU(),
    nn.Conv2d(384, 384, kernel_size=3, padding=1), nn.ReLU(),
    nn.Conv2d(384, 256, kernel_size=3, padding=1), nn.ReLU(),
    nn.MaxPool2d(kernel_size=3, stride=2), nn.Flatten(),
    nn.Linear(6400, 4096), nn.ReLU(), nn.Dropout(p=0.5),
    nn.Linear(4096, 4096), nn.ReLU(), nn.Dropout(p=0.5),
    nn.Linear(4096, 10))
```

AlexNet in Tensorflow

```
return tf.keras.models.Sequential([
   tf.keras.layers.Conv2D(filters=96, kernel_size=11, strides=4,
                          activation='relu')
   tf.keras.layers.MaxPool2D(pool_size=3, strides=2),
   tf.keras.layers.Conv2D(filters=256, kernel_size=5, padding='same',
                          activation='relu'),
   tf.keras.layers.MaxPool2D(pool_size=3, strides=2),
   tf.keras.layers.Conv2D(filters=384, kernel_size=3, padding='same',
                          activation='relu'),
   tf.keras.layers.Conv2D(filters=384, kernel_size=3, padding='same',
                          activation='relu'),
   tf.keras.layers.Conv2D(filters=256, kernel_size=3, padding='same',
                          activation='relu'),
   tf.keras.layers.MaxPool2D(pool_size=3, strides=2),
   tf.keras.layers.Flatten(),
   tf.keras.layers.Dense(4096, activation='relu'),
   tf.keras.layers.Dropout(0.5),
   tf.keras.layers.Dense(4096, activation='relu'),
   tf.keras.layers.Dropout(0.5),
   tf.keras.layers.Dense(10)])
```

VGG

VGG Network in PyTorch

```
def vgg_block(num_convs, in_channels, out_channels):
   lavers = []
   for _ in range(num_convs):
       layers.append(
           nn.Conv2d(in_channels, out_channels, kernel_size=3, padding=1))
       layers.append(nn.ReLU())
       in_channels = out_channels
   layers.append(nn.MaxPool2d(kernel_size=2, stride=2))
   return nn.Sequential(*layers)
def vgg(conv_arch):
   conv_blks = []
   in_channels = 1
   for (num_convs, out_channels) in conv_arch:
       conv_blks.append(vgg_block(num_convs, in_channels, out_channels))
       in_channels = out_channels
   return nn.Sequential(
       *conv_blks, nn.Flatten(),
       nn.Linear(out_channels * 7 * 7, 4096), nn.ReLU(), nn.Dropout(0.5),
       nn.Linear(4096, 4096), nn.ReLU(), nn.Dropout(0.5),
       nn.Linear(4096, 10))
conv_arch = ((1, 64), (1, 128), (2, 256), (2, 512), (2, 512))
net = vgg(conv_arch)
```

VGG Network in Tensorflow

```
def vgg_block(num_convs, num_channels):
   blk = tf.keras.models.Sequential()
   for _ in range(num_convs):
       blk.add(
            tf.keras.layers.Conv2D(num_channels, kernel_size=3,
                                   padding='same', activation='relu'))
   blk.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
   return blk
  def vgg(conv_arch):
   net = tf.keras.models.Sequential()
   for (num_convs, num_channels) in conv_arch:
        net.add(vgg_block(num_convs, num_channels))
   net.add(
        tf.keras.models.Sequential([
            tf.keras.layers.Flatten(),
            tf.keras.layers.Dense(4096, activation='relu'),
            tf.keras.layers.Dropout(0.5),
            tf.keras.layers.Dense(4096, activation='relu'),
            tf.keras.layers.Dropout(0.5),
            tf.keras.layers.Dense(10)]))
conv_arch = ((1, 64), (1, 128), (2, 256), (2, 512), (2, 512))
net = vgg(conv_arch)
```