CNN on MNIST

```
In [2]:
```

```
from __future__ import print_function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
from keras.layers.normalization import BatchNormalization
Using TensorFlow backend.
```

In [3]:

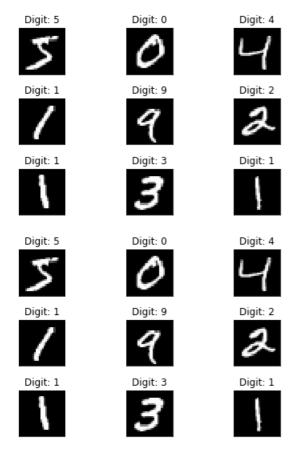
```
import matplotlib.pyplot as plt

(x_train, y_train), (x_test, y_test) = mnist.load_data()

fig = plt.figure()

for i in range(9):
    plt.subplot(3,3,i+1)
    plt.tight_layout()
    plt.imshow(x_train[i], cmap='gray', interpolation='none')
    plt.title("Digit: {}".format(y_train[i]))
    plt.xticks([])
    plt.yticks([])
```

Out[3]:



CNN + ReLU + ADAM with 3 Layers

In [4]:

```
from future import print function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
from keras.layers.normalization import BatchNormalization
batch size = 128
num classes = 10
epochs = 7
# input image dimensions
img rows, img cols = 28, 28
# the data, split between train and test sets
(x train, y train), (x test, y test) = mnist.load data()
if K.image data format() == 'channels first':
    x_train = x_train.reshape(x_train.shape[0], 1, img_rows, img_cols)
    x test = x_test.reshape(x_test.shape[0], 1, img_rows, img_cols)
   input shape = (1, img rows, img cols)
else:
    x train = x train.reshape(x train.shape[0], img rows, img cols, 1)
    x_test = x_test.reshape(x_test.shape[0], img_rows, img_cols, 1)
    input_shape = (img_rows, img_cols, 1)
x train = x train.astype('float32')
x_test = x_test.astype('float32')
x train /= 255
x_test /= 255
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x test.shape[0], 'test samples')
# convert class vectors to binary class matrices
y train = keras.utils.to categorical(y train, num classes)
y test = keras.utils.to categorical(y test, num classes)
model = Sequential()
model.add(Conv2D(24, kernel size=(7, 7),
                 activation='relu',
                 input shape=input_shape))
model.add(BatchNormalization())
model.add(Conv2D(32, (7, 7), activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(64, (7, 7), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.5))
model.add(Dense(num classes, activation='softmax'))
model.compile(loss=keras.losses.categorical crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])
model info=model.fit(x train, y train,
         batch size=batch size,
          epochs=epochs,
          verbose=1,
         validation data=(x test, y test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
x train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
```

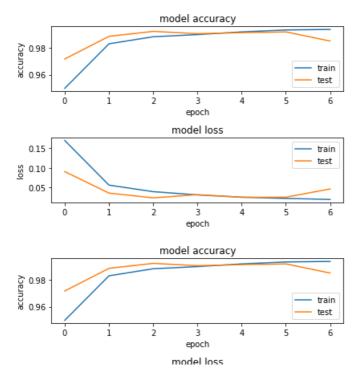
Train on 60000 samples, validate on 10000 samples

```
Epocn 1//
60000/60000 [==============] - 182s 3ms/step - loss: 0.1694 - acc: 0.9498 - val lo
ss: 0.0906 - val acc: 0.9717
Epoch 2/7
60000/60000 [=============] - 172s 3ms/step - loss: 0.0563 - acc: 0.9830 - val lo
ss: 0.0360 - val acc: 0.9886
Epoch 3/7
60000/60000 [=============] - 174s 3ms/step - loss: 0.0396 - acc: 0.9883 - val lo
ss: 0.0239 - val acc: 0.9923
Epoch 4/7
60000/60000 [=============] - 176s 3ms/step - loss: 0.0315 - acc: 0.9899 - val lo
ss: 0.0320 - val_acc: 0.9906
Epoch 5/7
60000/60000 [==============] - 151s 3ms/step - loss: 0.0256 - acc: 0.9919 - val lo
ss: 0.0254 - val acc: 0.9913
Epoch 6/7
60000/60000 [==============] - 177s 3ms/step - loss: 0.0225 - acc: 0.9933 - val lo
ss: 0.0255 - val_acc: 0.9919
Epoch 7/7
60000/60000 [=============] - 178s 3ms/step - loss: 0.0199 - acc: 0.9938 - val lo
ss: 0.0464 - val acc: 0.9851
Test loss: 0.046434765651240016
Test accuracy: 0.9851
```

In [5]:

```
from keras.models import Sequential
import os
import matplotlib.pyplot as plt
# plotting the metrics
fig = plt.figure()
plt.subplot(2,1,1)
plt.plot(model_info.history['acc'])
plt.plot(model_info.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='lower right')
plt.subplot(2,1,2)
plt.plot(model_info.history['loss'])
plt.plot(model info.history['val loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.tight layout()
fig
```

Out[5]:



In []:

CNN + ReLU + ADAM with 5 Layers

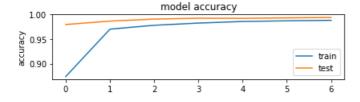
In [6]:

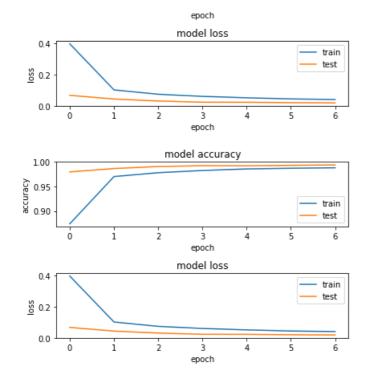
```
# Credits: https://github.com/keras-team/keras/blob/master/examples/mnist cnn.py
      _future__ import print_function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
batch size = 128
num classes = 10
epochs = 7
# input image dimensions
img rows, img cols = 28, 28
# the data, split between train and test sets
(x_train, y_train), (x_test, y_test) = mnist.load_data()
if K.image data format() == 'channels_first':
    x_train = x_train.reshape(x_train.shape[0], 1, img_rows, img_cols)
    x_test = x_test.reshape(x_test.shape[0], 1, img_rows, img_cols)
    input_shape = (1, img_rows, img_cols)
else:
    x_train = x_train.reshape(x_train.shape[0], img_rows, img_cols, 1)
    x test = x test.reshape(x test.shape[0], img rows, img cols, 1)
    input_shape = (img_rows, img_cols, 1)
x train = x train.astype('float32')
x_test = x_test.astype('float32')
x train /= 255
x test /= 255
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x_test.shape[0], 'test samples')
# convert class vectors to binary class matrices
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
model = Sequential()
model.add(Conv2D(24, kernel size=(3, 3),
                 activation='relu',
                 input shape=input shape))
model.add(Conv2D(32, (3, 3), activation='relu'))
model.add(Conv2D(32, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Dropout(0.25))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num classes, activation='softmax'))
```

```
model.compile(loss=keras.losses.categorical crossentropy,
           optimizer=keras.optimizers.Adadelta(),
            metrics=['accuracy'])
model info=model.fit(x_train, y_train,
        batch size=batch size,
        epochs=epochs,
        verbose=1,
        validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
x train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
Train on 60000 samples, validate on 10000 samples
Epoch 1/7
oss: 0.0674 - val acc: 0.9795
Epoch 2/7
60000/60000 [============== ] - 47s 785us/step - loss: 0.1019 - acc: 0.9700 - val 1
oss: 0.0440 - val acc: 0.9864
Epoch 3/7
oss: 0.0313 - val acc: 0.9905
Epoch 4/7
60000/60000 [============= ] - 53s 880us/step - loss: 0.0608 - acc: 0.9824 - val 1
oss: 0.0237 - val acc: 0.9922
Epoch 5/7
60000/60000 [============== ] - 50s 841us/step - loss: 0.0517 - acc: 0.9856 - val 1
oss: 0.0233 - val acc: 0.9919
Epoch 6/7
60000/60000 [=================== ] - 48s 801us/step - loss: 0.0447 - acc: 0.9869 - val 1
oss: 0.0208 - val acc: 0.9929
Epoch 7/7
60000/60000 [============== ] - 48s 792us/step - loss: 0.0409 - acc: 0.9879 - val 1
oss: 0.0196 - val acc: 0.9938
Test loss: 0.019573996598121448
Test accuracy: 0.9938
In [7]:
from keras.models import Sequential
import os
import matplotlib.pyplot as plt
# plotting the metrics
fig = plt.figure()
plt.subplot(2,1,1)
plt.plot(model_info.history['acc'])
plt.plot(model_info.history['val_acc'])
```

```
import os
import matplotlib.pyplot as plt
# plotting the metrics
fig = plt.figure()
plt.subplot(2,1,1)
plt.plot(model_info.history['acc'])
plt.plot(model_info.history['val_acc'])
plt.title('model accuracy')
plt.tylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='lower right')
plt.subplot(2,1,2)
plt.plot(model_info.history['loss'])
plt.plot(model_info.history['val_loss'])
plt.title('model loss')
plt.xlabel('epoch')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.legend(['train', 'test'], loc='upper right')
plt.tight_layout()
fig
```

Out[7]:





CNN + ReLU + ADAM with 7 Layers

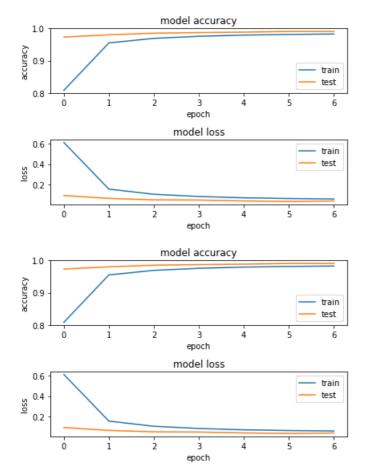
```
In [8]:
```

```
# Credits: https://github.com/keras-team/keras/blob/master/examples/mnist cnn.py
from future import print function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
from keras.layers.normalization import BatchNormalization
batch size = 128
num classes = 10
epochs = 7
# input image dimensions
img rows, img cols = 28, 28
# the data, split between train and test sets
(x train, y train), (x test, y test) = mnist.load data()
if K.image data format() == 'channels first':
    x_train = x_train.reshape(x_train.shape[0], 1, img_rows, img_cols)
    x_test = x_test.reshape(x_test.shape[0], 1, img_rows, img_cols)
    input_shape = (1, img_rows, img_cols)
else:
    x_train = x_train.reshape(x_train.shape[0], img_rows, img_cols, 1)
    x test = x test.reshape(x test.shape[0], img rows, img cols, 1)
    input_shape = (img_rows, img_cols, 1)
x train = x train.astype('float32')
x_test = x_test.astype('float32')
x_train /= 255
x_test /= 255
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x_test.shape[0], 'test samples')
# convert class vectors to binary class matrices
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
model = Sequential()
```

```
model.add(Conv2D(24, kernel size=(2, 2),
                activation='relu',
                input shape=input shape))
model.add(BatchNormalization())
model.add(Conv2D(32, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(32, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Dropout(0.25))
model.add(Conv2D(32, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(64, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Dropout(0.25))
model.add(Conv2D(64, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(Conv2D(64, (2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num classes, activation='softmax'))
model.compile(loss=keras.losses.categorical crossentropy,
             optimizer=keras.optimizers.Adadelta(),
             metrics=['accuracy'])
model info=model.fit(x train, y train,
         batch size=batch size,
         epochs=epochs,
         verbose=1,
         validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
x train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
Train on 60000 samples, validate on 10000 samples
60000/60000 [=============] - 193s 3ms/step - loss: 0.6113 - acc: 0.8086 - val lo
ss: 0.0943 - val acc: 0.9730
Epoch 2/7
60000/60000 [============= ] - 243s 4ms/step - loss: 0.1568 - acc: 0.9550 - val lo
ss: 0.0659 - val acc: 0.9800
Epoch 3/7
60000/60000 [============ ] - 195s 3ms/step - loss: 0.1063 - acc: 0.9690 - val lo
ss: 0.0519 - val acc: 0.9850
Epoch 4/7
ss: 0.0503 - val acc: 0.9868
Epoch 5/7
60000/60000 [=============] - 189s 3ms/step - loss: 0.0716 - acc: 0.9791 - val lo
ss: 0.0412 - val acc: 0.9881
Epoch 6/7
60000/60000 [=============] - 223s 4ms/step - loss: 0.0646 - acc: 0.9811 - val lo
ss: 0.0365 - val_acc: 0.9904
Epoch 7/7
60000/60000 [=============== ] - 244s 4ms/step - loss: 0.0604 - acc: 0.9825 - val lo
ss: 0.0403 - val acc: 0.9905
Test loss: 0.040273379761958494
Test accuracy: 0.9905
```

```
from keras.models import Sequential
import os
import matplotlib.pyplot as plt
# plotting the metrics
fig = plt.figure()
plt.subplot(2,1,1)
plt.plot(model info.history['acc'])
plt.plot(model_info.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='lower right')
plt.subplot(2,1,2)
plt.plot(model_info.history['loss'])
plt.plot(model_info.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.tight layout()
```

Out[9]:



Results

```
In [11]:
```

```
from prettytable import PrettyTable
x = PrettyTable()

names = ["Relu with 3 layers", "Relu with 5 layers", "Relu with 7 layers"]
Test_Accuracy = [98.51, 99.3, 99.05]
Test_Loss = [0.046,0.019,0.040]
numbering = [1,2,3]
ptable = PrettyTable()
# Adding columns
ptable.add_column("S.NO.", numbering)
ptable.add_column("MODEL", names)
ptable.add_column("Test_Score".Test_Loss)
```

```
ptable.add_column("Test Accuracy", Test_Accuracy)
# Printing the Table
print(ptable)
```

S.NO.		MODEL				-	Test Score	∍ +	Test Accuracy
1		Rel	u with	3	layers	i	0.046	i	98.51
2		Rel	u with	5	layers		0.019		99.3
3		Rel	u with	7	layers		0.04		99.05

In []: