

In [1]:

```
import os
os.environ["CUDA_DEVICE_ORDER"]="PCI_BUS_ID"
os.environ["CUDA_VISIBLE_DEVICES"]="{}".format(0)
```

In [2]:

```
import warnings
warnings.filterwarnings('ignore')
```

In [3]:

```
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
import matplotlib.pyplot as plt
```

Using TensorFlow backend.

Hyper parameters

In [4]:

```
batch_size = 128
num_classes = 10
epochs = 12
```

In [5]:

```
# 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()
```

In [6]:

```
print(K.image_data_format())
```

channels_last

In [7]:

```
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)
```

In [8]:

```
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)
```

```
x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
```

In [9]:

```
print(x_train.shape)
print(y_train.shape)
```

```
(60000, 28, 28, 1)
(60000, 10)
```

In [10]:

```
print(x_test.shape)
print(y_test.shape)
```

```
(10000, 28, 28, 1)
(10000, 10)
```

In [11]:

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
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.summary()
```

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 26, 26, 32)	320
conv2d_2 (Conv2D)	(None, 24, 24, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 12, 12, 64)	0
dropout_1 (Dropout)	(None, 12, 12, 64)	0
flatten_1 (Flatten)	(None, 9216)	0
dense_1 (Dense)	(None, 128)	1179776
dropout_2 (Dropout)	(None, 128)	0
dense_2 (Dense)	(None, 10)	1290
Total params: 1,199,882		
Trainable params: 1,199,882		
Non-trainable params: 0		

In [12]:

```
model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])
```

Keras callback 함수 이용하기

In [13]:

```
#from tensorflow.keras.callbacks import Callback
from keras.callbacks import Callback
from keras import backend as K

vloss = []
vacc = []
class NBatchLogger(Callback):

    def __init__(self, display):
        #self.step = 0
        self.display = display
        #self.metric_cache = {}

    #epoch 마다 learning rate 값 출력
    def on_epoch_end(self, epoch, logs=None):
        if self.display==1:
            print('aaaaa')
            global vloss
            global vacc

            vloss.append(logs['loss'])
            vacc.append(logs['acc'])
```

In [14]:

```
nbatch_logging = NBatchLogger(display=1)
```

In [15]:

```
model.fit(x_train, y_train,
          batch_size=batch_size,
          epochs=epochs,
          verbose=1,
          validation_data=(x_test, y_test),
          callbacks=[nbatch_logging])
```

Train on 60000 samples, validate on 10000 samples

Epoch 1/12

59392/60000 [=====>.] - ETA: 0s - loss: 0.3218
- acc: 0.9013aaaaa

60000/60000 [=====] - 7s 113us/step - loss:
0.3198 - acc: 0.9019 - val_loss: 0.0760 - val_acc: 0.9764

Epoch 2/12

59392/60000 [=====>.] - ETA: 0s - loss: 0.1117
- acc: 0.9666aaaaa

60000/60000 [=====] - 4s 64us/step - loss: 0.
1118 - acc: 0.9666 - val_loss: 0.0520 - val_acc: 0.9823

Epoch 3/12

59648/60000 [=====>.] - ETA: 0s - loss: 0.0882
- acc: 0.9744aaaaa

60000/60000 [=====] - 4s 65us/step - loss: 0.
0880 - acc: 0.9745 - val_loss: 0.0438 - val_acc: 0.9853

Epoch 4/12

59520/60000 [=====>.] - ETA: 0s - loss: 0.0712
- acc: 0.9784aaaaa

60000/60000 [=====] - 4s 64us/step - loss: 0.
0711 - acc: 0.9785 - val_loss: 0.0436 - val_acc: 0.9860

Epoch 5/12

59392/60000 [=====>.] - ETA: 0s - loss: 0.0629
- acc: 0.9818aaaaa

60000/60000 [=====] - 4s 64us/step - loss: 0.
0626 - acc: 0.9819 - val_loss: 0.0377 - val_acc: 0.9877

Epoch 6/12

59264/60000 [=====>.] - ETA: 0s - loss: 0.0551
- acc: 0.9838aaaaa

60000/60000 [=====] - 4s 65us/step - loss: 0.
0553 - acc: 0.9838 - val_loss: 0.0350 - val_acc: 0.9875

Epoch 7/12

59136/60000 [=====>.] - ETA: 0s - loss: 0.0505
- acc: 0.9845aaaaa

60000/60000 [=====] - 4s 65us/step - loss: 0.
0505 - acc: 0.9846 - val_loss: 0.0326 - val_acc: 0.9894

Epoch 8/12

59648/60000 [=====>.] - ETA: 0s - loss: 0.0468
- acc: 0.9863aaaaa

60000/60000 [=====] - 4s 65us/step - loss: 0.
0468 - acc: 0.9863 - val_loss: 0.0313 - val_acc: 0.9896

Epoch 9/12

59648/60000 [=====>.] - ETA: 0s - loss: 0.0442
- acc: 0.9866aaaaa

60000/60000 [=====] - 4s 65us/step - loss: 0.
0442 - acc: 0.9866 - val_loss: 0.0297 - val_acc: 0.9901

Epoch 10/12

59136/60000 [=====>.] - ETA: 0s - loss: 0.0421
- acc: 0.9875aaaaa

```
60000/60000 [=====] - 4s 65us/step - loss: 0.0420 - acc: 0.9875 - val_loss: 0.0322 - val_acc: 0.9899
Epoch 11/12
59904/60000 [=====>.] - ETA: 0s - loss: 0.0387 - acc: 0.9883aaaaaa
60000/60000 [=====] - 4s 65us/step - loss: 0.0387 - acc: 0.9883 - val_loss: 0.0293 - val_acc: 0.9901
Epoch 12/12
59904/60000 [=====>.] - ETA: 0s - loss: 0.0351 - acc: 0.9891aaaaaa
60000/60000 [=====] - 4s 65us/step - loss: 0.0351 - acc: 0.9891 - val_loss: 0.0294 - val_acc: 0.9897
```

Out[15]:

<keras.callbacks.History at 0x7f4e63c23080>

In [16]:

```
score = model.evaluate(x_train, y_train, verbose=0)
print('Train loss:', score[0])
print('Train accuracy:', score[1])
```

```
Train loss: 0.015571206274513194
Train accuracy: 0.99535
```

In [17]:

```
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

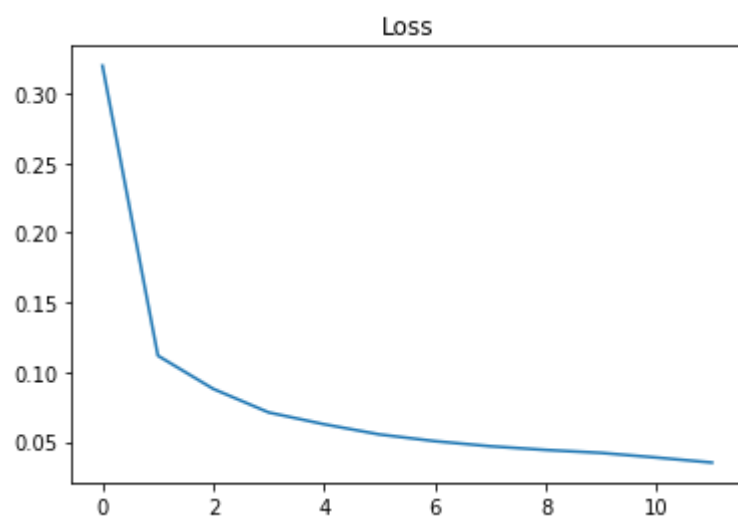
```
Test loss: 0.02944926852975186
Test accuracy: 0.9897
```

In [18]:

```
plt.plot(vloss)
plt.title('Loss')
```

Out[18]:

Text(0.5,1,'Loss')

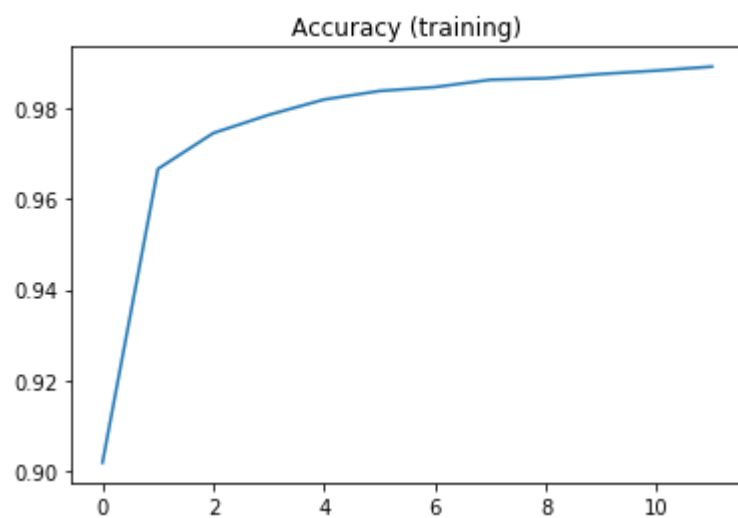


In [19]:

```
plt.plot(vacc)
plt.title('Accuracy (training)')
```

Out[19]:

Text(0.5,1,'Accuracy (training)')



Reference

- MNIST_CNN https://keras.io/examples/mnist_cnn/ (https://keras.io/examples/mnist_cnn/).

