### In [1]:

```
# 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
Using TensorFlow backend.
```

#### In [0]:

```
%matplotlib notebook
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import time
import pylab
# https://gist.github.com/greydanus/f6eee59eaf1d90fcb3b534a25362cea4
# https://stackoverflow.com/a/14434334
# this function is used to update the plots for each epoch and error
def plt_dynamic(x, vy, ty, ax, colors=['b']):
   ax.plot(x, vy, 'b', label="Validation Loss")
   ax.plot(x, ty, 'r', label="Train Loss")
   plt.legend()
   plt.grid()
   plt.show()
   pylab.show()
    #fig.canvas.draw()
```

## In [0]:

```
from keras.layers.normalization import BatchNormalization
from keras.layers import Dropout
```

## In [2]:

```
batch size = 128
num classes = 10
epochs = 12
# 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)
    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)
```

### **3 CONVOLUTION LAYER**

## In [4]:

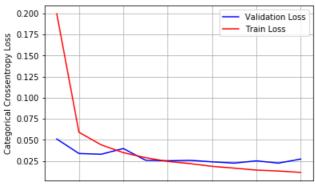
Layer (type)	Output	Shape	Param #
conv2d_3 (Conv2D)	(None,	24, 24, 100)	2600
conv2d_4 (Conv2D)	(None,	22, 22, 64)	57664
max_pooling2d_2 (MaxPooling2	(None,	11, 11, 64)	0
dropout_3 (Dropout)	(None,	11, 11, 64)	0
flatten_2 (Flatten)	(None,	7744)	0
dense_3 (Dense)	(None,	128)	991360
dropout_4 (Dropout)	(None,	128)	0
dense_4 (Dense)	(None,	10)	1290

Total params: 1,052,914
Trainable params: 1,052,914
Non-trainable params: 0

### In [5]

## In [6]:

```
print('Test accuracy:', score[1])
W0622 15:53:27.780707 140437071800192 deprecation.py:323| From /usr/local/lib/python3.6/dist-
packages/tensorflow/python/ops/math grad.py:1250: add dispatch support.<locals>.wrapper (from
tensorflow.python.ops.array_ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.where in 2.0, which has the same broadcast rule as np.where
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [============== ] - 334s 6ms/step - loss: 0.1996 - acc: 0.9381 - val lo
ss: 0.0509 - val acc: 0.9828
Epoch 2/12
60000/60000 [============== ] - 332s 6ms/step - loss: 0.0589 - acc: 0.9820 - val lo
ss: 0.0337 - val acc: 0.9881
Epoch 3/12
60000/60000 [============] - 332s 6ms/step - loss: 0.0440 - acc: 0.9867 - val lo
ss: 0.0328 - val_acc: 0.9895
Epoch 4/12
60000/60000 [============] - 332s 6ms/step - loss: 0.0347 - acc: 0.9893 - val lo
ss: 0.0396 - val_acc: 0.9857
Epoch 5/12
60000/60000 [============== ] - 332s 6ms/step - loss: 0.0288 - acc: 0.9912 - val lo
ss: 0.0258 - val_acc: 0.9918
Epoch 6/12
60000/60000 [============== ] - 332s 6ms/step - loss: 0.0243 - acc: 0.9923 - val lo
ss: 0.0251 - val acc: 0.9913
Epoch 7/12
60000/60000 [=============] - 332s 6ms/step - loss: 0.0218 - acc: 0.9928 - val lo
ss: 0.0257 - val acc: 0.9911
Epoch 8/12
60000/60000 [============== ] - 332s 6ms/step - loss: 0.0184 - acc: 0.9938 - val lo
ss: 0.0238 - val acc: 0.9929
Epoch 9/12
60000/60000 [============] - 332s 6ms/step - loss: 0.0164 - acc: 0.9947 - val lo
ss: 0.0223 - val acc: 0.9933
Epoch 10/12
60000/60000 [==============] - 332s 6ms/step - loss: 0.0140 - acc: 0.9953 - val lo
ss: 0.0250 - val acc: 0.9928
Epoch 11/12
60000/60000 [============= ] - 333s 6ms/step - loss: 0.0129 - acc: 0.9959 - val lo
ss: 0.0223 - val acc: 0.9934
Epoch 12/12
60000/60000 [=============== ] - 331s 6ms/step - loss: 0.0113 - acc: 0.9966 - val lo
ss: 0.0271 - val acc: 0.9919
Test loss: 0.027147230633624533
Test accuracy: 0.9919
In [15]:
fig,ax = plt.subplots(1,1)
ax.set xlabel('epoch') ; ax.set ylabel('Categorical Crossentropy Loss')
x = list(range(1, epochs+1))
vy = history.history['val loss']
ty = history.history['loss']
plt dynamic(x, vy, ty, ax)
```



verbose=1.

print('Test loss:', score[0])

validation\_data=(x\_test, y\_test))
score = model.evaluate(x test, y test, verbose=0)

## **5 CONVOLUTION LAYER**

### In [16]:

```
model1 = Sequential()
model1.add(Conv2D(100, kernel_size=(5, 5),
                  activation='relu',
                  input_shape=input_shape))
model1.add(Conv2D(64, (4, 4), activation='relu'))
model1.add(MaxPooling2D(pool_size=(2, 2)))
model1.add(Dropout(0.25))
model1.add(Conv2D(42, (3, 3), activation='relu'))
model1.add(MaxPooling2D(pool_size=(2, 2)))
model1.add(BatchNormalization())
model1.add(Conv2D(30, (2, 2), activation='relu'))
model1.add(Flatten())
model1.add(Dense(128, activation='relu'))
model1.add(Dropout(0.25))
model1.add(Dense(num_classes, activation='softmax'))
model1.summary()
W0622 17:21:45.065067 140437071800192 deprecation_wrapper.py:119] From
/usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow backend.py:1834: The name
tf.nn.fused batch norm is deprecated. Please use tf.compat.v1.nn.fused batch norm instead.
```

Layer (type)	Output	Shape	Param #
conv2d_5 (Conv2D)	(None,	24, 24, 100)	2600
conv2d_6 (Conv2D)	(None,	21, 21, 64)	102464
max_pooling2d_3 (MaxPooling2	(None,	10, 10, 64)	0
dropout_5 (Dropout)	(None,	10, 10, 64)	0
conv2d_7 (Conv2D)	(None,	8, 8, 42)	24234
max_pooling2d_4 (MaxPooling2	(None,	4, 4, 42)	0
<pre>batch_normalization_1 (Batch</pre>	(None,	4, 4, 42)	168
conv2d_8 (Conv2D)	(None,	3, 3, 30)	5070
flatten_3 (Flatten)	(None,	270)	0
dense_5 (Dense)	(None,	128)	34688
dropout_6 (Dropout)	(None,	128)	0
dense_6 (Dense)	(None,	10)	1290
Total params: 170,514 Trainable params: 170,430 Non-trainable params: 84			=======

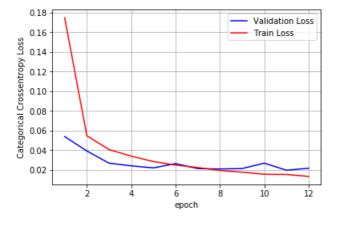
## In [0]:

# In [19]:

```
score = model1.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
Train on 60000 samples, validate on 10000 samples
60000/60000 [============= ] - 501s 8ms/step - loss: 0.1747 - acc: 0.9460 - val lo
ss: 0.0539 - val acc: 0.9840
Epoch 2/12
60000/60000 [============== ] - 501s 8ms/step - loss: 0.0548 - acc: 0.9837 - val lo
ss: 0.0393 - val acc: 0.9865
Epoch 3/12
60000/60000 [============== ] - 501s 8ms/step - loss: 0.0407 - acc: 0.9875 - val lo
ss: 0.0269 - val acc: 0.9916
Epoch 4/12
60000/60000 [============== ] - 502s 8ms/step - loss: 0.0340 - acc: 0.9892 - val lo
ss: 0.0242 - val acc: 0.9923
Epoch 5/12
60000/60000 [============= ] - 501s 8ms/step - loss: 0.0285 - acc: 0.9913 - val lo
ss: 0.0221 - val acc: 0.9925
Epoch 6/12
60000/60000 [=============] - 500s 8ms/step - loss: 0.0250 - acc: 0.9923 - val lo
ss: 0.0265 - val acc: 0.9907
Epoch 7/12
60000/60000 [=============] - 502s 8ms/step - loss: 0.0224 - acc: 0.9932 - val lo
ss: 0.0213 - val_acc: 0.9937
Epoch 8/12
60000/60000 [==============] - 501s 8ms/step - loss: 0.0195 - acc: 0.9940 - val lo
ss: 0.0210 - val acc: 0.9934
Epoch 9/12
60000/60000 [============== ] - 500s 8ms/step - loss: 0.0178 - acc: 0.9942 - val lo
ss: 0.0215 - val_acc: 0.9936
Epoch 10/12
60000/60000 [============== ] - 502s 8ms/step - loss: 0.0156 - acc: 0.9952 - val lo
ss: 0.0269 - val_acc: 0.9916
Epoch 11/12
60000/60000 [============== ] - 503s 8ms/step - loss: 0.0154 - acc: 0.9953 - val lo
ss: 0.0197 - val acc: 0.9940
Epoch 12/12
60000/60000 [============= ] - 503s 8ms/step - loss: 0.0134 - acc: 0.9959 - val lo
ss: 0.0218 - val acc: 0.9938
Test loss: 0.021787987656591713
Test accuracy: 0.9938
```

## In [20]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
vy = history1.history['val_loss']
ty = history1.history['loss']
plt_dynamic(x, vy, ty, ax)
```



validation data=(x test, y test))

### 7 LAYER CNN

```
model2 = Sequential()
model2.add(Conv2D(100, kernel_size=(8, 8),
                 activation='relu',
                 input_shape=input_shape))
model2.add(Conv2D(90, (7, 7), activation='relu'))
#model2.add(MaxPooling2D(pool_size=(5, 5)))
model2.add(Dropout(0.25))
model2.add(Conv2D(80, (6, 6), activation='relu'))
#model2.add(MaxPooling2D(pool size=(3, 3)))
model2.add(BatchNormalization())
model2.add(Conv2D(30, (4, 4), activation='relu'))
#model2.add(MaxPooling2D(pool size=(2, 2)))
model2.add(Conv2D(25, (3, 3), activation='relu'))
model2.add(Dropout(0.1))
model2.add(BatchNormalization())
model2.add(Conv2D(20, (2, 2), activation='relu'))
model2.add(Dropout(0.15))
model2.add(BatchNormalization())
model2.add(Flatten())
model2.add(Dense(128, activation='relu'))
model2.add(Dropout(0.25))
model2.add(Dense(num classes, activation='softmax'))
model2.summary()
```

Layer (type)	Output Shape	Param #
conv2d_62 (Conv2D)	(None, 21, 21, 100)	6500
conv2d_63 (Conv2D)	(None, 15, 15, 90)	441090
dropout_28 (Dropout)	(None, 15, 15, 90)	0
conv2d_64 (Conv2D)	(None, 10, 10, 80)	259280
batch_normalization_15 (Batc	(None, 10, 10, 80)	320
conv2d_65 (Conv2D)	(None, 7, 7, 30)	38430
conv2d_66 (Conv2D)	(None, 5, 5, 25)	6775
dropout_29 (Dropout)	(None, 5, 5, 25)	0
batch_normalization_16 (Batc	(None, 5, 5, 25)	100
conv2d_67 (Conv2D)	(None, 4, 4, 20)	2020
dropout_30 (Dropout)	(None, 4, 4, 20)	0
batch_normalization_17 (Batc	(None, 4, 4, 20)	80
flatten_8 (Flatten)	(None, 320)	0
dense_15 (Dense)	(None, 128)	41088
dropout_31 (Dropout)	(None, 128)	0
dense_16 (Dense)	(None, 10)	1290
Total params: 796,973 Trainable params: 796,723 Non-trainable params: 250		=

In [0]:

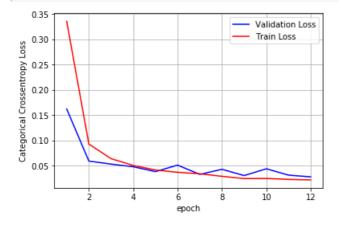
## In [37]:

```
verbose=1,
       validation data=(x test, y test))
score = model2.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
loss: 0.1625 - val acc: 0.9509
Epoch 2/12
60000/60000 [=============] - 1311s 22ms/step - loss: 0.0929 - acc: 0.9732 - val
loss: 0.0592 - val acc: 0.9828
Epoch 3/12
loss: 0.0533 - val acc: 0.9845
Epoch 4/12
60000/60000 [============= ] - 1310s 22ms/step - loss: 0.0506 - acc: 0.9850 - val
loss: 0.0480 - val_acc: 0.9861
Epoch 5/12
60000/60000 [============== ] - 1311s 22ms/step - loss: 0.0416 - acc: 0.9884 - val
loss: 0.0384 - val_acc: 0.9881
Epoch 6/12
60000/60000 [=============] - 1314s 22ms/step - loss: 0.0371 - acc: 0.9893 - val
loss: 0.0512 - val acc: 0.9840
Epoch 7/12
60000/60000 [============== ] - 1317s 22ms/step - loss: 0.0341 - acc: 0.9903 - val
loss: 0.0329 - val acc: 0.9895
Epoch 8/12
loss: 0.0430 - val acc: 0.9869
Epoch 9/12
loss: 0.0305 - val acc: 0.9917
Epoch 10/12
60000/60000 [============== ] - 1329s 22ms/step - loss: 0.0249 - acc: 0.9928 - val
loss: 0.0440 - val acc: 0.9882
Epoch 11/12
60000/60000 [============= ] - 1333s 22ms/step - loss: 0.0230 - acc: 0.9936 - val
loss: 0.0314 - val acc: 0.9903
Epoch 12/12
60000/60000 [============= ] - 1333s 22ms/step - loss: 0.0220 - acc: 0.9938 - val
loss: 0.0280 - val acc: 0.9920
Test loss: 0.028006677371503655
```

### In [38]:

Test accuracy: 0.992

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch'); ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
vy = history2.history['val_loss']
ty = history2.history['loss']
plt_dynamic(x, vy, ty, ax)
```



epochs=epochs,

+----+