CNN on MNIST dataset

In [1]:
```
from tensorflow.python.client import device_lib
print(device_lib.list_local_devices())
```

```
[name: "/device:CPU:0"
device_type: "CPU"
memory_limit: 268435456
locality {
}
incarnation: 16988170116324900273
, name: "/device:GPU:0"
device_type: "GPU"
memory_limit: 3206820659
locality {
  bus_id: 1
  links {
  }
}
incarnation: 199497037252222683
physical_device_desc: "device: 0, name: GeForce 940MX, pci bus id: 000
0:01:00.0, compute capability: 5.0"
]
```

In [2]:
```
from __future__ import print_function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Conv2D, MaxPool2D, Flatten, Dropout, Ba
tchNormalization
from keras import backend as K
import matplotlib.pyplot as plt
```

```
Using TensorFlow backend.
```

In [3]:
```
%matplotlib notebook
import matplotlib.pyplot as plt
import numpy as np
import time
# 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()
    fig.canvas.draw()
```

In [4]:
```
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()
```

In [5]:
```
print(K.image_data_format())
print(x_train.shape)
```

```
channels_last
(60000, 28, 28)
```

In [6]:
```
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)
```

In [7]:
```
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, 1, 28, 28)
60000 train samples
10000 test samples
```

3 conv Layered model

In [ ]:
```
model3.reset_states()
```

In [37]:
```
model3= Sequential()

model3.add(Conv2D(32,kernel_size=3,activation='relu',input_shape=input_
shape,data_format='channels_first'))
model3.add(BatchNormalization(axis=-1)) #for conv layer axis=1 ,,

model3.add(Conv2D(160,kernel_size=4,activation='relu'))
model3.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model3.add(Dropout(0.3))
model3.add(MaxPool2D(pool_size=2))

model3.add(Conv2D(196,kernel_size=3,activation='relu'))
model3.add(BatchNormalization(axis=-1)) #for conv layer axis=1

model3.add(Flatten())
model3.add(Dense(128, activation='relu'))
model3.add(Dense(64, activation='relu'))
model3.add(Dropout(0.5))
model3.add(Dense(num_classes, activation='softmax'))
```

In [38]:
```
model3.summary()
```

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
conv2d_30 (Conv2D)           (None, 32, 26, 26)        320
_________________________________________________________________
batch_normalization_27 (Batc (None, 32, 26, 26)        104
_________________________________________________________________
conv2d_31 (Conv2D)           (None, 29, 23, 160)       66720
_________________________________________________________________
batch_normalization_28 (Batc (None, 29, 23, 160)       640
_________________________________________________________________
dropout_23 (Dropout)         (None, 29, 23, 160)       0
_________________________________________________________________
max_pooling2d_8 (MaxPooling2 (None, 14, 11, 160)       0
_________________________________________________________________
conv2d_32 (Conv2D)           (None, 12, 9, 196)        282436
_________________________________________________________________
batch_normalization_29 (Batc (None, 12, 9, 196)        784
_________________________________________________________________
flatten_5 (Flatten)          (None, 21168)             0
_________________________________________________________________
dense_13 (Dense)             (None, 128)               2709632
_________________________________________________________________
dense_14 (Dense)             (None, 64)                8256
_________________________________________________________________
dropout_24 (Dropout)         (None, 64)                0
_________________________________________________________________
dense_15 (Dense)             (None, 10)                650
=================================================================
Total params: 3,069,542
Trainable params: 3,068,778
Non-trainable params: 764
_________________________________________________________________
```

In [39]:
```
model3.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])
```

```
history3=  model3.fit(x_train, y_train,
          batch_size=128,
          epochs=10,
          verbose=1,
          validation_data=(x_test, y_test))
score = model3.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

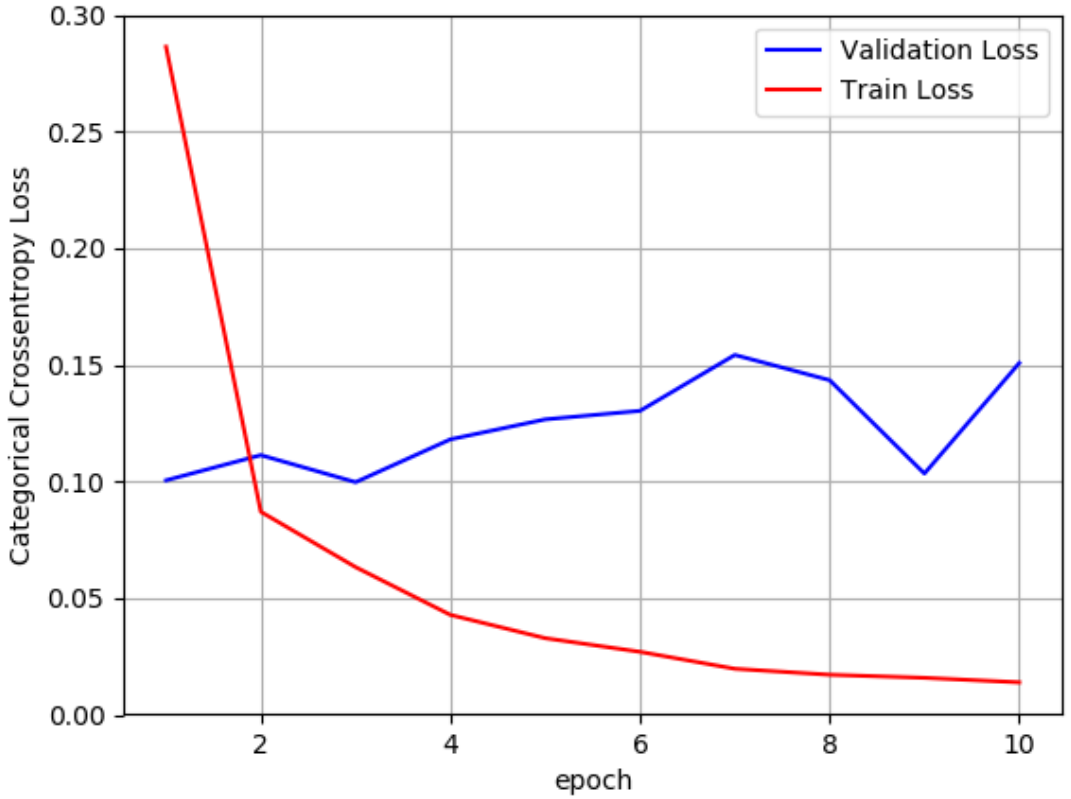
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/10
60000/60000 [==============================] - 266s 4ms/step - loss: 0.
2863 - acc: 0.9235 - val_loss: 0.1005 - val_acc: 0.9764
Epoch 2/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0870 - acc: 0.9789 - val_loss: 0.1113 - val_acc: 0.9780
Epoch 3/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0633 - acc: 0.9849 - val_loss: 0.0997 - val_acc: 0.9859
Epoch 4/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0428 - acc: 0.9893 - val_loss: 0.1181 - val_acc: 0.9848
Epoch 5/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0329 - acc: 0.9922 - val_loss: 0.1266 - val_acc: 0.9863
Epoch 6/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0270 - acc: 0.9935 - val_loss: 0.1303 - val_acc: 0.9856
Epoch 7/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0197 - acc: 0.9954 - val_loss: 0.1542 - val_acc: 0.9854
Epoch 8/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0172 - acc: 0.9961 - val_loss: 0.1434 - val_acc: 0.9860
Epoch 9/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
0159 - acc: 0.9961 - val_loss: 0.1034 - val_acc: 0.9897
Epoch 10/10
60000/60000 [==============================] - 261s 4ms/step - loss: 0.
```

```
0139 - acc: 0.9969 - val_loss: 0.1508 - val_acc: 0.9863
Test loss: 0.15075194108878912
Test accuracy: 0.9863
```

In [40]:
```
score= model3.evaluate(x_test, y_test, verbose=0)
print('Test score: ',score[0])
print('Test accuracy: ',score[1])
```

```
Test score:  0.15075194108878912
Test accuracy:  0.9863
```

In [41]:
```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,11))
vy = model3.history.history['val_loss']
ty = model3.history.history['loss']
plt_dynamic(x, vy, ty, ax)
```



In [42]:
```
pickling_on= open("Model3.pickle","wb")
pickle.dump(model3, pickling_on)
pickling_on.close()
```

4 Conv layered model

In [9]:
```
model= Sequential()

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

```
ut_shape,data_format='channels_first'))
model.add(BatchNormalization(axis=1)) #for conv layer axis=1 ,,

model.add(Conv2D(35,kernel_size=3,activation='relu'))
model.add(BatchNormalization(axis=1)) #for conv layer axis=1
model.add(MaxPool2D(pool_size=2))
model.add(Dropout(0.25))

model.add(Conv2D(70,kernel_size=3,activation='relu'))
model.add(BatchNormalization(axis=1)) #for conv layer axis=1
model.add(Dropout(0.25))

model.add(Conv2D(100,kernel_size=2,activation='relu'))
model.add(BatchNormalization(axis=1)) #for conv layer axis=1
model.add(MaxPool2D(pool_size=2))

model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))
```

```
WARNING:tensorflow:From C:\Anaconda3\lib\site-packages\tensorflow\pytho
n\framework\op_def_library.py:263: colocate_with (from tensorflow.pytho
n.framework.ops) is deprecated and will be removed in a future version.
Instructions for updating:
Colocations handled automatically by placer.
WARNING:tensorflow:From C:\Anaconda3\lib\site-packages\keras\backend\te
nsorflow_backend.py:3445: calling dropout (from tensorflow.python.ops.n
n_ops) with keep_prob is deprecated and will be removed in a future ver
sion.
Instructions for updating:
Please use `rate` instead of `keep_prob`. Rate should be set to `rate =
1 - keep_prob`.
```

In [10]:
```
model.summary()
```

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
```

```
conv2d_1 (Conv2D)            (None, 32, 26, 26)        320
_________________________________________________________________
batch_normalization_1 (Batch (None, 32, 26, 26)        128
_________________________________________________________________
conv2d_2 (Conv2D)            (None, 30, 24, 35)        8225
_________________________________________________________________
batch_normalization_2 (Batch (None, 30, 24, 35)        120
_________________________________________________________________
max_pooling2d_1 (MaxPooling2 (None, 15, 12, 35)        0
_________________________________________________________________
dropout_1 (Dropout)          (None, 15, 12, 35)        0
_________________________________________________________________
conv2d_3 (Conv2D)            (None, 13, 10, 70)        22120
_________________________________________________________________
batch_normalization_3 (Batch (None, 13, 10, 70)        52
_________________________________________________________________
dropout_2 (Dropout)          (None, 13, 10, 70)        0
_________________________________________________________________
conv2d_4 (Conv2D)            (None, 12, 9, 100)        28100
_________________________________________________________________
batch_normalization_4 (Batch (None, 12, 9, 100)        48
_________________________________________________________________
max_pooling2d_2 (MaxPooling2 (None, 6, 4, 100)         0
_________________________________________________________________
flatten_1 (Flatten)          (None, 2400)              0
_________________________________________________________________
dense_1 (Dense)              (None, 128)               307328
_________________________________________________________________
dense_2 (Dense)              (None, 64)                8256
_________________________________________________________________
dropout_3 (Dropout)          (None, 64)                0
_________________________________________________________________
dense_3 (Dense)              (None, 10)                650
=================================================================
Total params: 375,347
Trainable params: 375,173
Non-trainable params: 174
_________________________________________________________________
```

In [11]:
```
model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

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])
```

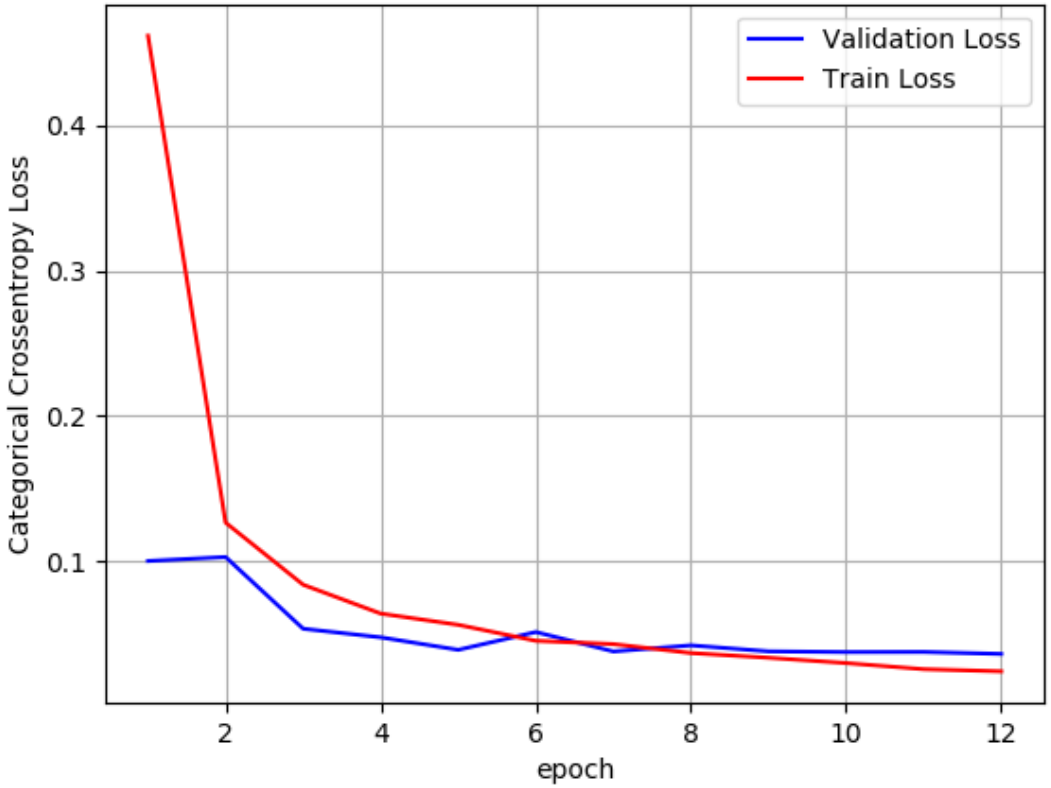
```
WARNING:tensorflow:From C:\Anaconda3\lib\site-packages\tensorflow\pytho
n\ops\math_ops.py:3066: to_int32 (from tensorflow.python.ops.math_ops)
is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [==============================] - 384s 6ms/step - loss: 0.
4619 - acc: 0.8613 - val_loss: 0.1001 - val_acc: 0.9718
Epoch 2/12
60000/60000 [==============================] - 365s 6ms/step - loss: 0.
1266 - acc: 0.9663 - val_loss: 0.1029 - val_acc: 0.9726
Epoch 3/12
60000/60000 [==============================] - 364s 6ms/step - loss: 0.
0838 - acc: 0.9779 - val_loss: 0.0535 - val_acc: 0.9867
Epoch 4/12
60000/60000 [==============================] - 381s 6ms/step - loss: 0.
0639 - acc: 0.9829 - val_loss: 0.0476 - val_acc: 0.9885
Epoch 5/12
60000/60000 [==============================] - 378s 6ms/step - loss: 0.
0562 - acc: 0.9859 - val_loss: 0.0389 - val_acc: 0.9889
Epoch 6/12
60000/60000 [==============================] - 381s 6ms/step - loss: 0.
0452 - acc: 0.9877 - val_loss: 0.0512 - val_acc: 0.9897
Epoch 7/12
60000/60000 [==============================] - 367s 6ms/step - loss: 0.
0429 - acc: 0.9885 - val_loss: 0.0378 - val_acc: 0.9905
```

```
Epoch 8/12
60000/60000 [==============================] - 367s 6ms/step - loss: 0.
0367 - acc: 0.9906 - val_loss: 0.0420 - val_acc: 0.9916
Epoch 9/12
60000/60000 [==============================] - 368s 6ms/step - loss: 0.
0335 - acc: 0.9910 - val_loss: 0.0379 - val_acc: 0.9922
Epoch 10/12
60000/60000 [==============================] - 367s 6ms/step - loss: 0.
0298 - acc: 0.9922 - val_loss: 0.0374 - val_acc: 0.9910
Epoch 11/12
60000/60000 [==============================] - 366s 6ms/step - loss: 0.
0256 - acc: 0.9934 - val_loss: 0.0375 - val_acc: 0.9924
Epoch 12/12
60000/60000 [==============================] - 370s 6ms/step - loss: 0.
0241 - acc: 0.9938 - val_loss: 0.0361 - val_acc: 0.9925
Test loss: 0.036131656656459746
Test accuracy: 0.9925
```

In [12]:
```
score= model.evaluate(x_test, y_test, verbose=0)
print('Test score: ',score[0])
print('Test accuracy: ',score[1])
```

```
Test score:  0.036131656656459746
Test accuracy:  0.9925
```

In [13]:
```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,13))
vy = model.history.history['val_loss']
ty = model.history.history['loss']
plt_dynamic(x, vy, ty, ax)
```



In [32]:
```
import pickle
pickling_on= open("Model1.pickle","wb")
pickle.dump(model, pickling_on)
pickling_on.close()
```

7 conv layered model

In [20]:
```
model2.reset_states()
```

In [21]:
```
model2= Sequential()

model2.add(Conv2D(32,kernel_size=6,activation='relu',input_shape=input_
shape,data_format='channels_first'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1 ,,

model2.add(Conv2D(150,kernel_size=5,activation='relu'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model2.add(Dropout(0.3))

model2.add(Conv2D(150,kernel_size=4,activation='relu'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model2.add(Dropout(0.2))

model2.add(Conv2D(200,kernel_size=4,padding='same',activation='relu'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model2.add(Dropout(0.25))

model2.add(Conv2D(200,kernel_size=3,activation='relu'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model2.add(Dropout(0.35))

model2.add(Conv2D(150,kernel_size=3,padding='same',activation='relu'))
model2.add(Dropout(0.25))

model2.add(Conv2D(180,kernel_size=3,activation='relu'))
model2.add(BatchNormalization(axis=-1)) #for conv layer axis=1
model2.add(MaxPool2D(pool_size=2))

model2.add(Flatten())
model2.add(Dense(128, activation='relu'))
model2.add(Dense(64, activation='relu'))
model2.add(Dropout(0.5))
model2.add(Dense(num_classes, activation='softmax'))
```

In [22]:
```
model2.summary()
```

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
```

```
=================================================================
conv2d_20 (Conv2D)           (None, 32, 23, 23)        1184
_________________________________________________________________
batch_normalization_18 (Batc (None, 32, 23, 23)        92
_________________________________________________________________
conv2d_21 (Conv2D)           (None, 28, 19, 150)       86400
_________________________________________________________________
batch_normalization_19 (Batc (None, 28, 19, 150)       600
_________________________________________________________________
dropout_15 (Dropout)         (None, 28, 19, 150)       0
_________________________________________________________________
conv2d_22 (Conv2D)           (None, 25, 16, 150)       360150
_________________________________________________________________
batch_normalization_20 (Batc (None, 25, 16, 150)       600
_________________________________________________________________
dropout_16 (Dropout)         (None, 25, 16, 150)       0
_________________________________________________________________
conv2d_23 (Conv2D)           (None, 25, 16, 200)       480200
_________________________________________________________________
batch_normalization_21 (Batc (None, 25, 16, 200)       800
_________________________________________________________________
dropout_17 (Dropout)         (None, 25, 16, 200)       0
_________________________________________________________________
conv2d_24 (Conv2D)           (None, 23, 14, 200)       360200
_________________________________________________________________
batch_normalization_22 (Batc (None, 23, 14, 200)       800
_________________________________________________________________
dropout_18 (Dropout)         (None, 23, 14, 200)       0
_________________________________________________________________
conv2d_25 (Conv2D)           (None, 23, 14, 150)       270150
_________________________________________________________________
dropout_19 (Dropout)         (None, 23, 14, 150)       0
_________________________________________________________________
conv2d_26 (Conv2D)           (None, 21, 12, 180)       243180
_________________________________________________________________
batch_normalization_23 (Batc (None, 21, 12, 180)       720
_________________________________________________________________
max_pooling2d_5 (MaxPooling2 (None, 10, 6, 180)        0
_________________________________________________________________
```

```
flatten_3 (Flatten)          (None, 10800)             0
_________________________________________________________________
dense_7 (Dense)              (None, 128)               1382528
_________________________________________________________________
dense_8 (Dense)              (None, 64)                8256
_________________________________________________________________
dropout_20 (Dropout)         (None, 64)                0
_________________________________________________________________
dense_9 (Dense)              (None, 10)                650
=================================================================
Total params: 3,196,510
Trainable params: 3,194,704
Non-trainable params: 1,806
_________________________________________________________________
```

In [23]:

```
model2.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

history2=  model2.fit(x_train, y_train,
          batch_size=150,
          epochs=13,
          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 1/13
60000/60000 [==============================] - 902s 15ms/step - loss:
0.5966 - acc: 0.8275 - val_loss: 0.2085 - val_acc: 0.9585
Epoch 2/13
60000/60000 [==============================] - 888s 15ms/step - loss:
0.1367 - acc: 0.9654 - val_loss: 0.0849 - val_acc: 0.9800
Epoch 3/13
60000/60000 [==============================] - 888s 15ms/step - loss:
0.0893 - acc: 0.9787 - val_loss: 0.0911 - val_acc: 0.9822
Epoch 4/13
```

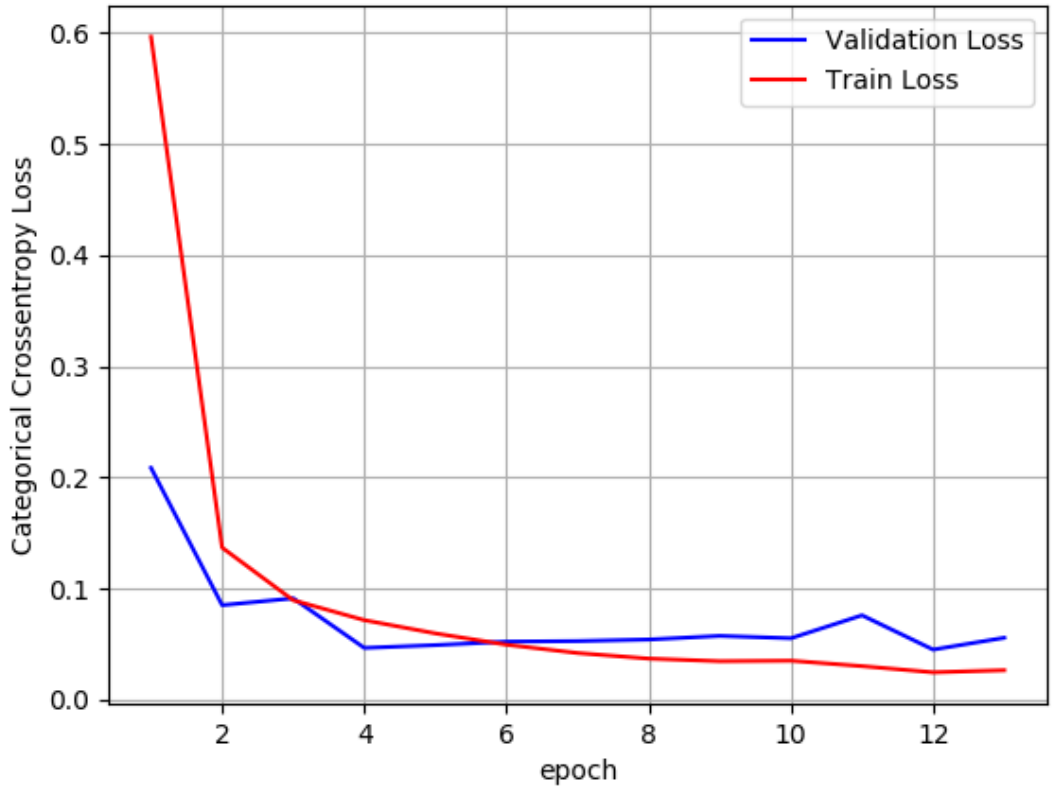
```
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0714 - acc: 0.9831 - val_loss: 0.0464 - val_acc: 0.9907
Epoch 5/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0595 - acc: 0.9858 - val_loss: 0.0490 - val_acc: 0.9900
Epoch 6/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0493 - acc: 0.9876 - val_loss: 0.0521 - val_acc: 0.9900
Epoch 7/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0418 - acc: 0.9897 - val_loss: 0.0526 - val_acc: 0.9908
Epoch 8/13
60000/60000 [==============================] - 888s 15ms/step - loss:
0.0368 - acc: 0.9907 - val_loss: 0.0540 - val_acc: 0.9905
Epoch 9/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0345 - acc: 0.9915 - val_loss: 0.0572 - val_acc: 0.9907
Epoch 10/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0349 - acc: 0.9921 - val_loss: 0.0551 - val_acc: 0.9907
Epoch 11/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0300 - acc: 0.9929 - val_loss: 0.0758 - val_acc: 0.9880
Epoch 12/13
60000/60000 [==============================] - 887s 15ms/step - loss:
0.0244 - acc: 0.9944 - val_loss: 0.0449 - val_acc: 0.9906
Epoch 13/13
60000/60000 [==============================] - 886s 15ms/step - loss:
0.0263 - acc: 0.9940 - val_loss: 0.0555 - val_acc: 0.9909
Test loss: 0.05553952041512905
Test accuracy: 0.9909
```

In [24]:

```
score= model2.evaluate(x_test, y_test, verbose=0)
print('Test score: ',score[0])
print('Test accuracy: ',score[1])
```

```
Test score:  0.05553952041512905
Test accuracy:  0.9909
```

In [25]:
```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,14))
vy = model2.history.history['val_loss']
ty = model2.history.history['loss']
plt_dynamic(x, vy, ty, ax)
```



In [29]:
```
import pickle
pickling_on= open("Model2.pickle","wb")
pickle.dump(model2, pickling_on)
pickling_on.close()
```

Summary

Model	# of conv layer	Accuracy
1	3	98.63
2	4	99.35
3	7	99.09