

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
```

Using TensorFlow backend.

In [0]:

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

In [0]:

```
img_rows, img_cols = 28, 28
```

In [5]:

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

Downloading data from <https://s3.amazonaws.com/img-datasets/mnist.npz>  
11493376/11490434 [=====] - 1s 0us/step

In [0]:

```
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 [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')
```

x\_train shape: (60000, 28, 28, 1)  
60000 train samples  
10000 test samples

In [0]:

```
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
```

In [0]:

```
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()
```

### Model 1: 3 convolutional layer

In [0]:

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu', input_shape=input_shape))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(64, kernel_size=(3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(128, kernel_size=(3, 3), activation='relu'))

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'))
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/framework/op\_def\_library.py:263: colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version. Instructions for updating:  
Colocations handled automatically by placer.  
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3445: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version. Instructions for updating:  
Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

In [0]:

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

In [0]:

```
history = 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])
```

Train on 60000 samples, validate on 10000 samples  
Epoch 1/12  
60000/60000 [=====] - 68s 1ms/step - loss: 0.2760 - acc: 0.9161 - val\_loss: 0.0895 - val\_acc: 0.9708  
Epoch 2/12  
60000/60000 [=====] - 69s 1ms/step - loss: 0.1426 - acc: 0.9589 - val\_loss: 0.0571 - val\_acc: 0.9821  
Epoch 3/12  
60000/60000 [=====] - 69s 1ms/step - loss: 0.1132 - acc: 0.9672 - val\_loss: 0.0556 - val\_acc: 0.9845  
Epoch 4/12  
60000/60000 [=====] - 69s 1ms/step - loss: 0.0998 - acc: 0.9723 - val\_loss: 0.0484 - val\_acc: 0.9857  
Epoch 5/12  
60000/60000 [=====] - 69s 1ms/step - loss: 0.0900 - acc: 0.9747 - val\_loss: 0.0362 - val\_acc: 0.9889  
Epoch 6/12  
60000/60000 [=====] - 69s 1ms/step - loss: 0.0824 - acc: 0.9767 - val\_loss: 0.0475 - val\_acc: 0.9870  
Epoch 7/12  
60000/60000 [=====] - 70s 1ms/step - loss: 0.0770 - acc: 0.9790 - val\_loss:

```

s: 0.0466 - val_acc: 0.9865
Epoch 8/12
60000/60000 [=====] - 70s 1ms/step - loss: 0.0725 - acc: 0.9806 - val_loss: 0.0487 - val_acc: 0.9862
Epoch 9/12
60000/60000 [=====] - 69s 1ms/step - loss: 0.0722 - acc: 0.9809 - val_loss: 0.0418 - val_acc: 0.9885
Epoch 10/12
60000/60000 [=====] - 73s 1ms/step - loss: 0.0687 - acc: 0.9817 - val_loss: 0.0365 - val_acc: 0.9907
Epoch 11/12
60000/60000 [=====] - 70s 1ms/step - loss: 0.0654 - acc: 0.9831 - val_loss: 0.0345 - val_acc: 0.9910
Epoch 12/12
60000/60000 [=====] - 70s 1ms/step - loss: 0.0672 - acc: 0.9830 - val_loss: 0.0390 - val_acc: 0.9888
Test loss: 0.03899886542790646
Test accuracy: 0.9888

```

In [0]:

```

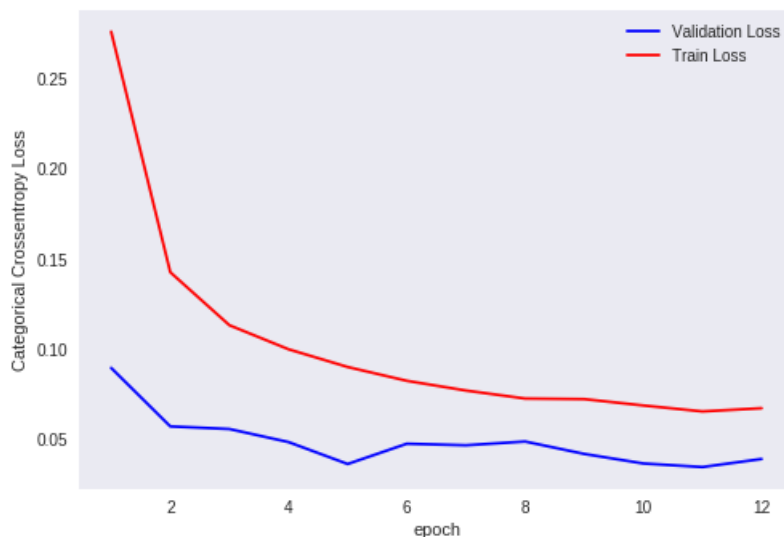
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])

fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')

# list of epoch numbers
x = list(range(1, epochs+1))
vy = history.history['val_loss']
ty = history.history['loss']
plt_dynamic(x, vy, ty, ax)

```

Test loss: 0.03899886542790646  
Test accuracy: 0.9888



Observation: As number of epochs increases overfitting reduces.

Model 2: 5 convolutional layer

In [0]:

```

from keras.layers import BatchNormalization

```

In [11]:

```

model = Sequential()
model.add(Conv2D(32, kernel_size=(2, 2), activation='relu', input_shape=input_shape))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(2, 2), activation='relu'))

```

```

model.add(Conv2D(32, kernel_size=(2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(128, kernel_size=(2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(2, 2), activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Flatten())
model.add(Dense(10, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/framework/op\_def\_library.py:263: colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.  
Instructions for updating:  
Colocations handled automatically by placer.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3445: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version.  
Instructions for updating:  
Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

In [0]:

```

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

```

In [13]:

```

history = 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 /usr/local/lib/python3.6/dist-packages/tensorflow/python/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 [=====] - 21s 358us/step - loss: 1.3492 - acc: 0.5311 - val\_loss: 0.4087 - val\_acc: 0.9483

Epoch 2/12  
60000/60000 [=====] - 18s 296us/step - loss: 0.9433 - acc: 0.6766 - val\_loss: 0.2985 - val\_acc: 0.9671

Epoch 3/12  
60000/60000 [=====] - 18s 297us/step - loss: 0.8652 - acc: 0.6960 - val\_loss: 0.2397 - val\_acc: 0.9789

Epoch 4/12  
60000/60000 [=====] - 18s 295us/step - loss: 0.8294 - acc: 0.7091 - val\_loss: 0.2097 - val\_acc: 0.9818

Epoch 5/12  
60000/60000 [=====] - 18s 298us/step - loss: 0.7820 - acc: 0.7207 - val\_loss: 0.2032 - val\_acc: 0.9808

```

Epoch 6/12
60000/60000 [=====] - 18s 297us/step - loss: 0.7705 - acc: 0.7257 - val_loss: 0.1752 - val_acc: 0.9863
Epoch 7/12
60000/60000 [=====] - 18s 298us/step - loss: 0.7443 - acc: 0.7357 - val_loss: 0.1701 - val_acc: 0.9868
Epoch 8/12
60000/60000 [=====] - 18s 297us/step - loss: 0.7285 - acc: 0.7381 - val_loss: 0.1601 - val_acc: 0.9870
Epoch 9/12
60000/60000 [=====] - 18s 296us/step - loss: 0.7289 - acc: 0.7382 - val_loss: 0.1646 - val_acc: 0.9860
Epoch 10/12
60000/60000 [=====] - 18s 299us/step - loss: 0.7271 - acc: 0.7392 - val_loss: 0.1605 - val_acc: 0.9870
Epoch 11/12
60000/60000 [=====] - 18s 299us/step - loss: 0.7176 - acc: 0.7411 - val_loss: 0.1547 - val_acc: 0.9863
Epoch 12/12
60000/60000 [=====] - 18s 298us/step - loss: 0.7161 - acc: 0.7418 - val_loss: 0.1523 - val_acc: 0.9886
Test loss: 0.15232067085504533
Test accuracy: 0.9886

```

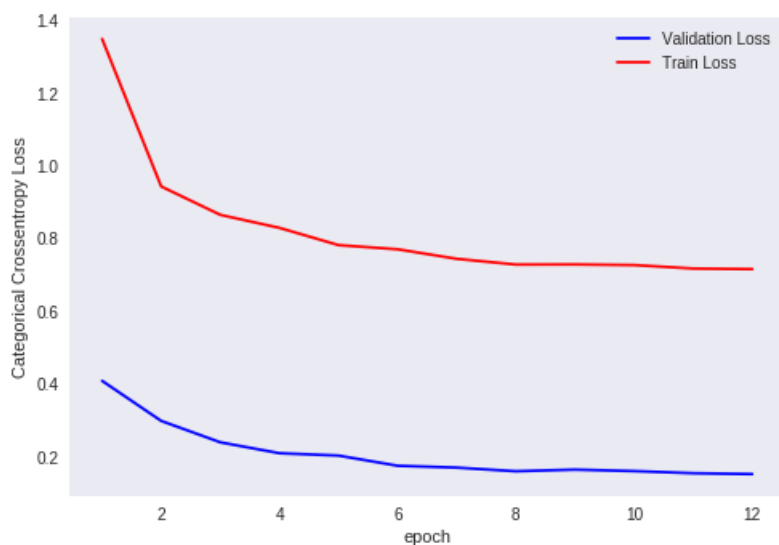
In [15]:

```

import matplotlib.pyplot as plt
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')

# list of epoch numbers
x = list(range(1,epochs+1))
vy = history.history['val_loss']
ty = history.history['loss']
plt_dynamic(x, vy, ty, ax)

```



Observation: As number of epochs increases overfitting not that much reduces as in 3 convolutional layer.

Model 3: 7convolutional layer

In [0]:

```

from keras.layers import BatchNormalization
model = Sequential()
model.add(Conv2D(32, kernel_size=(2,2),activation='relu',input_shape=input_shape))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(128, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))

```

```

model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

model.add(Conv2D(512, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(512, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(128, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(2,2),activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Flatten())
model.add(Dense(10, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

```

In [0]:

```

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

```

In [18]:

```

history = 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])

```

Train on 60000 samples, validate on 10000 samples

```

Epoch 1/12
60000/60000 [=====] - 81s 1ms/step - loss: 1.2123 - acc: 0.5546 - val_loss: 0.2543 - val_acc: 0.9397
Epoch 2/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.8338 - acc: 0.6634 - val_loss: 0.1111 - val_acc: 0.9762
Epoch 3/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.7547 - acc: 0.6901 - val_loss: 0.0831 - val_acc: 0.9825
Epoch 4/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.7196 - acc: 0.7046 - val_loss: 0.0635 - val_acc: 0.9868
Epoch 5/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6941 - acc: 0.7220 - val_loss: 0.0632 - val_acc: 0.9874
Epoch 6/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6742 - acc: 0.7358 - val_loss: 0.0671 - val_acc: 0.9892
Epoch 7/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6613 - acc: 0.7451 - val_loss: 0.0665 - val_acc: 0.9872
Epoch 8/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6409 - acc: 0.7548 - val_loss: 0.0715 - val_acc: 0.9885
Epoch 9/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6385 - acc: 0.7573 - val_loss: 0.0717 - val_acc: 0.9866
Epoch 10/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6399 - acc: 0.7582 - val_loss: 0.0549 - val_acc: 0.9894
Epoch 11/12

```

```

60000/60000 [=====] - 78s 1ms/step - loss: 0.6261 - acc: 0.7657 - val_loss: 0.0603 - val_acc: 0.9897
Epoch 12/12
60000/60000 [=====] - 78s 1ms/step - loss: 0.6189 - acc: 0.7696 - val_loss: 0.0623 - val_acc: 0.9892
Test loss: 0.062323937116935846
Test accuracy: 0.9892

```

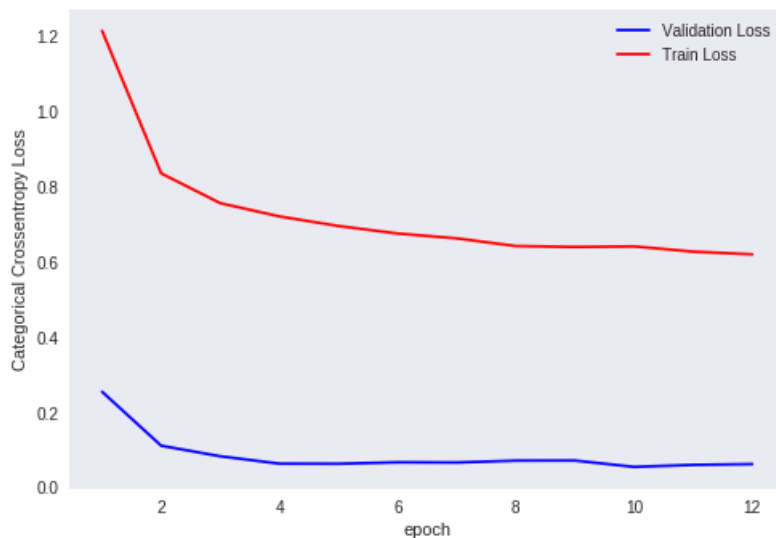
In [19]:

```

fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')

# list of epoch numbers
x = list(range(1,epochs+1))
vy = history.history['val_loss']
ty = history.history['loss']
plt_dynamic(x, vy, ty, ax)

```



Observation: As number of epochs increases overfitting not that much reduces as in 3 convolutional layer.

### Conclusion:

In [21]:

```

from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Model", "Test AUC", "Test Loss"]

x.add_row(["Model 1: 3 convolutional layer", "98.88%", "0.03899886542790646"])
x.add_row(["Model 2: 5 convolutional layer", "98.86%", "0.15232067085504533"])
x.add_row(["Model 3: 7 convolutional layer", "98.92%", "0.062323937116935846"])

print(x)

```

Model	Test AUC	Test Loss
Model 1: 3 convolutional layer	98.88%	0.03899886542790646
Model 2: 5 convolutional layer	98.86%	0.15232067085504533
Model 3: 7 convolutional layer	98.92%	0.062323937116935846

1) As number of layers increases accuracy increases. 2) Test loss very loss in 3 convolutional layer.

In [0]:

