```
In [21]:
# 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, BatchNormalization
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
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)
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)
x train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
In [10]:
import matplotlib.pyplot as plt
%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")
```

- 1. CN-->Conv2d
- 2. MP-->max\_pooling2d

plt.legend()
plt.grid()
fig.canvas.draw()

- 3. DP-->Dropout
- 4. FL-->Flatten
- 5. DN-->Dense

# 1st Model CN(72,(7,7))->CN(38,(7,7))->CN(20,(7,7))->MP(2,2)->DP(0.25)->FL()->DN(10)

#### In [29]:

```
K.clear_session()
model = Sequential()
model.add(Conv2D(72, kernel_size=(7, 7),activation='relu',input_shape=input_shape))
model.add(Conv2D(38, (7, 7), activation='relu'))
model.add(Conv2D(20, (7, 7), 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, 22, 22, 72)	3600
conv2d_2 (Conv2D)	(None, 16, 16, 38)	134102
conv2d_3 (Conv2D)	(None, 10, 10, 20)	37260
max_pooling2d_1 (MaxPooling2	(None, 5, 5, 20)	0
dropout_1 (Dropout)	(None, 5, 5, 20)	0
flatten_1 (Flatten)	(None, 500)	0
dense_1 (Dense)	(None, 128)	64128
dropout_2 (Dropout)	(None, 128)	0
dense_2 (Dense)	(None, 10)	1290
Total params: 240,380 Trainable params: 240,380 Non-trainable params: 0		

#### In [14]:

model.compile(loss=keras.losses.categorical\_crossentropy,optimizer='adam', metrics=['accuracy'])
history=model.fit(x\_train, y\_train,batch\_size=batch\_size,epochs=epochs,verbose=1,validation\_data=(x\_test, y\_test))

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [============= ] - 66s 1ms/step - loss: 0.2616 - acc: 0.9211 - val los
s: 0.0515 - val acc: 0.9839
Epoch 2/12
60000/60000 [============== ] - 64s 1ms/step - loss: 0.0912 - acc: 0.9725 - val_los
s: 0.0402 - val acc: 0.9860
Epoch 3/12
60000/60000 [=============] - 64s 1ms/step - loss: 0.0694 - acc: 0.9790 - val los
s: 0.0330 - val acc: 0.9893
Epoch 4/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0582 - acc: 0.9824 - val los
s: 0.0326 - val acc: 0.9893
Epoch 5/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0504 - acc: 0.9852 - val los
s: 0.0278 - val acc: 0.9913
Epoch 6/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0474 - acc: 0.9864 - val los
s: 0.0296 - val acc: 0.9897
Epoch 7/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0413 - acc: 0.9873 - val los
```

```
s: 0.0260 - val acc: 0.9910
Epoch 8/12
60000/60000 [============== ] - 65s 1ms/step - loss: 0.0362 - acc: 0.9888 - val los
s: 0.0239 - val acc: 0.9928
Epoch 9/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0360 - acc: 0.9888 - val los
s: 0.0258 - val_acc: 0.9920
Epoch 10/12
60000/60000 [============ ] - 65s 1ms/step - loss: 0.0333 - acc: 0.9898 - val los
s: 0.0240 - val_acc: 0.9925
Epoch 11/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0301 - acc: 0.9902 - val los
s: 0.0217 - val acc: 0.9930
Epoch 12/12
60000/60000 [============= ] - 65s 1ms/step - loss: 0.0283 - acc: 0.9910 - val los
s: 0.0234 - val acc: 0.9926
```

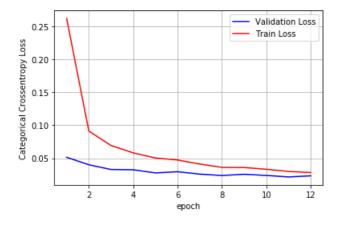
#### In [15]:

```
score = model.evaluate(x_test, y_test, verbose=0)
print('Test score:', 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 score: 0.023416508061222523 Test accuracy: 0.9926



## 2nd Model CN(20,(5,5))->BN()->CN(20)->BN()->CN(40)->BN()->CN(10)->BN()->CN(10)->MP(2,2)->DP(0.25)->FL()->DN(128)

#### In [22]:

```
K.clear session()
model_1 = Sequential()
model 1.add(Conv2D(20, kernel size=(5, 5),activation='relu',input shape=input shape))
model 1.add(BatchNormalization())
model 1.add(Conv2D(20, (5, 5), activation='relu'))
model 1.add(BatchNormalization())
model_1.add(Conv2D(40, (5, 5), activation='relu'))
model_1.add(BatchNormalization())
model 1.add(Conv2D(10, (5, 5), activation='relu'))
model 1.add(BatchNormalization())
model 1.add(Conv2D(10, (5, 5), activation='relu'))
model_1.add(MaxPooling2D(pool_size=(2, 2)))
model_1.add(Dropout(0.25))
model_1.add(Flatten())
model 1.add(Dense(128, activation='relu'))
model 1.add(Dropout(0.5))
```

```
model_1.add(Dense(num_classes, activation='softmax'))
model_1.summary()
```

WARNING:tensorflow:From C:\Users\nnagari\AppData\Local\Continuum\anaconda3\envs\tf\_gpu\lib\site-pa ckages\keras\backend\tensorflow\_backend.py:1834: The name tf.nn.fused\_batch\_norm is deprecated. Pl ease use tf.compat.v1.nn.fused\_batch\_norm instead.

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 24, 24, 20)	520
batch_normalization_1 (Batch	(None, 24, 24, 20)	80
conv2d_2 (Conv2D)	(None, 20, 20, 20)	10020
batch_normalization_2 (Batch	(None, 20, 20, 20)	80
conv2d_3 (Conv2D)	(None, 16, 16, 40)	20040
batch_normalization_3 (Batch	(None, 16, 16, 40)	160
conv2d_4 (Conv2D)	(None, 12, 12, 10)	10010
batch_normalization_4 (Batch	(None, 12, 12, 10)	40
conv2d_5 (Conv2D)	(None, 8, 8, 10)	2510
max_pooling2d_1 (MaxPooling2	(None, 4, 4, 10)	0
dropout_1 (Dropout)	(None, 4, 4, 10)	0
flatten_1 (Flatten)	(None, 160)	0
dense_1 (Dense)	(None, 128)	20608
dropout_2 (Dropout)	(None, 128)	0
dense_2 (Dense)	(None, 10)	1290

Total params: 65,358 Trainable params: 65,178 Non-trainable params: 180

#### In [23]:

model\_1.compile(loss=keras.losses.categorical\_crossentropy,optimizer='adam', metrics=['accuracy'])
history=model\_1.fit(x\_train, y\_train,batch\_size=batch\_size,epochs=epochs,verbose=1,validation\_data=
(x\_test, y\_test))

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [=============] - 52s 870us/step - loss: 0.4221 - acc: 0.8676 - val 1
oss: 0.0828 - val acc: 0.9767
Epoch 2/12
60000/60000 [============== ] - 50s 833us/step - loss: 0.1077 - acc: 0.9700 - val 1
oss: 0.0470 - val_acc: 0.9852
Epoch 3/12
60000/60000 [============== ] - 50s 831us/step - loss: 0.0775 - acc: 0.9775 - val 1
oss: 0.0450 - val acc: 0.9863
Epoch 4/12
60000/60000 [============= ] - 50s 832us/step - loss: 0.0664 - acc: 0.9813 - val 1
oss: 0.0429 - val_acc: 0.9867
Epoch 5/12
60000/60000 [============= ] - 50s 834us/step - loss: 0.0562 - acc: 0.9835 - val 1
oss: 0.0338 - val acc: 0.9903
Epoch 6/12
60000/60000 [============= ] - 50s 832us/step - loss: 0.0494 - acc: 0.9858 - val 1
oss: 0.0301 - val_acc: 0.9908
Epoch 7/12
60000/60000 [============== ] - 50s 834us/step - loss: 0.0482 - acc: 0.9861 - val 1
oss: 0.0300 - val_acc: 0.9922
60000/60000 [============== ] - 50s 834us/step - loss: 0.0462 - acc: 0.9865 - val 1
```

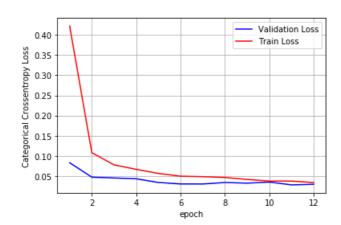
#### In [25]:

```
score = model_1.evaluate(x_test, y_test, verbose=0)
print('Test score:', 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 score: 0.029317251708592812 Test accuracy: 0.9914



### $3rd\ Model\ CN(30,(5,5))->BN()->CN(30)->BN()->CN(50)->BN()->CN(100)->CN(100)->BN()->CN(100)->C$

#### In [44]:

```
K.clear session()
model 2 = Sequential()
model 2.add(Conv2D(30, kernel size=(3, 3),activation='relu',input shape=input shape))
model_2.add(BatchNormalization())
model_2.add(Conv2D(30, (3, 3), activation='relu'))
model 2.add(BatchNormalization())
model 2.add(Conv2D(50, (3, 3), activation='relu'))
model 2.add(BatchNormalization())
model 2.add(Conv2D(50, (3, 3), activation='relu'))
model_2.add(BatchNormalization())
model_2.add(Conv2D(100, (3, 3), activation='relu'))
model 2.add(BatchNormalization())
model_2.add(Conv2D(100, (3, 3), activation='relu'))
model 2.add(MaxPooling2D(pool size=(2, 2)))
model 2.add(BatchNormalization())
model_2.add(Conv2D(150, (3, 3), activation='relu'))
model_2.add(MaxPooling2D(pool_size=(2, 2)))
model 2.add(Dropout(0.25))
model 2.add(Flatten())
```

```
model_2.add(Dense(120, activation='relu'))
model_2.add(Dropout(0.5))
model_2.add(Dense(num_classes, activation='softmax'))
model_2.summary()
```

Layer (type)	Output	Shape	Param #
conv2d_1 (Conv2D)	(None,	26, 26, 30)	300
batch_normalization_1 (Batch	(None,	26, 26, 30)	120
conv2d_2 (Conv2D)	(None,	24, 24, 30)	8130
batch_normalization_2 (Batch	(None,	24, 24, 30)	120
conv2d_3 (Conv2D)	(None,	22, 22, 50)	13550
batch_normalization_3 (Batch	(None,	22, 22, 50)	200
conv2d_4 (Conv2D)	(None,	20, 20, 50)	22550
batch_normalization_4 (Batch	(None,	20, 20, 50)	200
conv2d_5 (Conv2D)	(None,	18, 18, 100)	45100
<pre>batch_normalization_5 (Batch</pre>	(None,	18, 18, 100)	400
conv2d_6 (Conv2D)	(None,	16, 16, 100)	90100
max_pooling2d_1 (MaxPooling2	(None,	8, 8, 100)	0
batch_normalization_6 (Batch	(None,	8, 8, 100)	400
conv2d_7 (Conv2D)	(None,	6, 6, 150)	135150
max_pooling2d_2 (MaxPooling2	(None,	3, 3, 150)	0
dropout_1 (Dropout)	(None,	3, 3, 150)	0
flatten_1 (Flatten)	(None,	1350)	0
dense_1 (Dense)	(None,	120)	162120
dropout_2 (Dropout)	(None,	120)	0
dense_2 (Dense)	(None,	10)	1210

Total params: 479,650 Trainable params: 478,930 Non-trainable params: 720

#### In [45]:

model\_2.compile(loss=keras.losses.categorical\_crossentropy,optimizer='adam', metrics=['accuracy'])
history=model\_2.fit(x\_train, y\_train,batch\_size=batch\_size,epochs=epochs,verbose=1,validation\_data=
(x\_test, y\_test))

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [==============] - 195s 3ms/step - loss: 0.2658 - acc: 0.9202 - val lo
ss: 0.0506 - val acc: 0.9843
Epoch 2/12
ss: 0.0383 - val acc: 0.9891
Epoch 3/12
60000/60000 [============== ] - 198s 3ms/step - loss: 0.0750 - acc: 0.9799 - val lo
ss: 0.0309 - val_acc: 0.9903
Epoch 4/12
60000/60000 [============= ] - 197s 3ms/step - loss: 0.0681 - acc: 0.9812 - val lo
ss: 0.0270 - val_acc: 0.9915
Epoch 5/12
60000/60000 [============= ] - 198s 3ms/step - loss: 0.0546 - acc: 0.9852 - val lo
ss: 0.0338 - val acc: 0.9901
```

```
Epoch 6/12
60000/60000 [==============] - 198s 3ms/step - loss: 0.0450 - acc: 0.9872 - val lo
ss: 0.0300 - val acc: 0.9924
Epoch 7/12
ss: 0.0251 - val acc: 0.9911
Epoch 8/12
60000/60000 [==============] - 198s 3ms/step - loss: 0.0402 - acc: 0.9886 - val lo
ss: 0.0353 - val acc: 0.9898
Epoch 9/12
60000/60000 [=============] - 198s 3ms/step - loss: 0.0386 - acc: 0.9893 - val lo
ss: 0.0234 - val_acc: 0.9940
Epoch 10/12
60000/60000 [=============] - 199s 3ms/step - loss: 0.0361 - acc: 0.9902 - val lo
ss: 0.0260 - val_acc: 0.9925
Epoch 11/12
60000/60000 [==============] - 198s 3ms/step - loss: 0.0347 - acc: 0.9906 - val lo
ss: 0.0278 - val acc: 0.9916
Epoch 12/12
60000/60000 [=============] - 197s 3ms/step - loss: 0.0295 - acc: 0.9917 - val lo
ss: 0.0343 - val acc: 0.9923
```

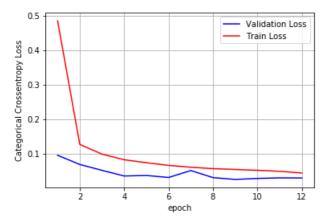
#### In [41]:

```
score = model_2.evaluate(x_test, y_test, verbose=0)
print('Test score:', 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 score: 0.029114628794253077 Test accuracy: 0.9915



#### Observation

We have 28X28 size image if we add max pooling layer multiple times in Network demission of the image will reduce and we will get following error "Negative dimension size caused by subtracting 3 from 2 for 'conv2d\_7/convolution' (op: 'Conv2D') with input shapes: [?,2,2,20], [3,3,20,20]."

#### In [59]:

```
print("(Model 1st:-", "CN(72,(7,7))->CN(38,(7,7))->CN(20,(7,7))->MP(2,2)->DP(0.25)->FL()->DN(10)\
n")
print("(Model 2nd:-CN(20,(5,5))->BN()->CN(20)->BN()->CN(40)->BN()->CN(10)->BN()->CN(10)->BN()->CN(10)->MP(2,2)->
DP(0.25)->FL()->DN(128)\n")
print("(Model 3rd:-", "CN(30,(5,5))->BN()->CN(30)->BN()->CN(50)->BN()->CN(50)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(100)->BN()->CN(
```

```
(\texttt{Model 1st:- CN}(72, (7,7)) -> \texttt{CN}(38, (7,7)) -> \texttt{CN}(20, (7,7)) -> \texttt{MP}(2,2) -> \texttt{DP}(0.25) -> \texttt{FL}() -> \texttt{DN}(10))
 (\mathsf{Model}\ 2\mathsf{nd} : -\mathsf{CN}(20, (5,5)) -> \mathsf{BN}() -> \mathsf{CN}(20) -> \mathsf{BN}() -> \mathsf{CN}(40) -> \mathsf{BN}() -> \mathsf{CN}(10) -> \mathsf{CN}(
>DP(0.25)->FL()->DN(128)
 (Model 3rd:- CN(30,(5,5))->BN()->CN(30)->BN()->CN(50)->BN()->CN(50)->BN()->CN(100)->BN()->CN(100)-
>MP(2,2)->BN()->CN(150)->MP(2,2)->DP(0.25)->FL()->DN(120)
In [62]:
 from prettytable import PrettyTable
 table = PrettyTable()
 table.field names = ["Model", "Accuracy"]
 table.add row(["Model 1st","0.9926"])
 table.add_row(["Model 2nd","0.9914"])
 table.add row(["Model 3nd","0.9915"])
print(table)
| Model | Accuracy |
 +----+
 | Model 1st | 0.9926 |
 | Model 2nd | 0.9914 |
| Model 3nd | 0.9915 |
```