

CNN ON MNIST

1. 2 CONVOLUTIONAL LAYERS (MAX POOLING AND DROPOUT INCLUDED)

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

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)

model1 = Sequential()
model1.add(Conv2D(32, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D(pool_size=(2, 2)))
model1.add(Dropout(0.25))
model1.add(Flatten())
model1.add(Dense(128, activation='relu'))
model1.add(Dropout(0.5))
model1.add(Dense(num_classes, activation='softmax'))

model1.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model1.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])

```

Using TensorFlow backend.

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

x_train shape: (60000, 28, 28, 1)

60000 train samples

10000 test samples

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:66: The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4432: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4267: The name tf.nn.max_pool is deprecated. Please use tf.nn.max_pool2d instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:148: The name tf.placeholder_with_default is deprecated. Please use tf.compat.v1.placeholder_with_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3733: 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`.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

WARNING:tensorflow: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 [=====] - 16s 260us/step - loss: 0.2620 - acc: 0.9203 - val_loss: 0.0547 - val_acc: 0.9829

Epoch 2/12

60000/60000 [=====] - 9s 144us/step - loss: 0.0875 - acc: 0.9735 - val_loss: 0.0379 - val_acc: 0.9863

Epoch 3/12

60000/60000 [=====] - 9s 144us/step - loss: 0.0652 - acc: 0.9803 - val_loss: 0.0325 - val_acc: 0.9890

Epoch 4/12

60000/60000 [=====] - 9s 144us/step - loss: 0.0550 - acc: 0.9834 - val_loss: 0.0351 - val_acc: 0.9884

Epoch 5/12

60000/60000 [=====] - 9s 144us/step - loss: 0.0478 - acc: 0.9855 - val_loss: 0.0276 - val_acc: 0.9908

Epoch 6/12

60000/60000 [=====] - 9s 143us/step - loss: 0.0417 - acc: 0.9875 - val_loss: 0.0288 - val_acc: 0.9901

Epoch 7/12

60000/60000 [=====] - 9s 144us/step - loss: 0.0380 - acc: 0.9881 - val_loss: 0.0220 - val_acc: 0.9905

```
- val_loss: 0.0329 - val_acc: 0.9885
Epoch 8/12
60000/60000 [=====] - 9s 144us/step - loss: 0.0335 - acc: 0.9895
- val_loss: 0.0333 - val_acc: 0.9895
Epoch 9/12
60000/60000 [=====] - 9s 143us/step - loss: 0.0318 - acc: 0.9900
- val_loss: 0.0267 - val_acc: 0.9924
Epoch 10/12
60000/60000 [=====] - 9s 144us/step - loss: 0.0306 - acc: 0.9907
- val_loss: 0.0268 - val_acc: 0.9916
Epoch 11/12
60000/60000 [=====] - 9s 146us/step - loss: 0.0287 - acc: 0.9916
- val_loss: 0.0272 - val_acc: 0.9921
Epoch 12/12
60000/60000 [=====] - 9s 145us/step - loss: 0.0272 - acc: 0.9915
- val_loss: 0.0267 - val_acc: 0.9922
Test loss: 0.026746683429056065
Test accuracy: 0.9922
```

In [0]:

```
model1.summary()
```

| Layer (type) | Output Shape | Param # |
|--------------------------------|--------------------|---------|
| conv2d_9 (Conv2D) | (None, 26, 26, 32) | 320 |
| conv2d_10 (Conv2D) | (None, 24, 24, 64) | 18496 |
| max_pooling2d_6 (MaxPooling2D) | (None, 12, 12, 64) | 0 |
| dropout_8 (Dropout) | (None, 12, 12, 64) | 0 |
| flatten_5 (Flatten) | (None, 9216) | 0 |
| dense_9 (Dense) | (None, 128) | 1179776 |
| dropout_9 (Dropout) | (None, 128) | 0 |
| dense_10 (Dense) | (None, 10) | 1290 |
| Total params: 1,199,882 | | |
| Trainable params: 1,199,882 | | |
| Non-trainable params: 0 | | |

In [0]:

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import time
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 [0]:

```
def plt_dynamic1(x, vy, ty, ax, colors=['b']):
    ax.plot(x, vy, 'b', label="Validation Accuracy")
    ax.plot(x, ty, 'r', label="Training Accuracy")
    plt.legend()
    plt.grid()
    fig.canvas.draw()
```

In [0]:

```
score = model1.evaluate(x test, y test, verbose=0)
```

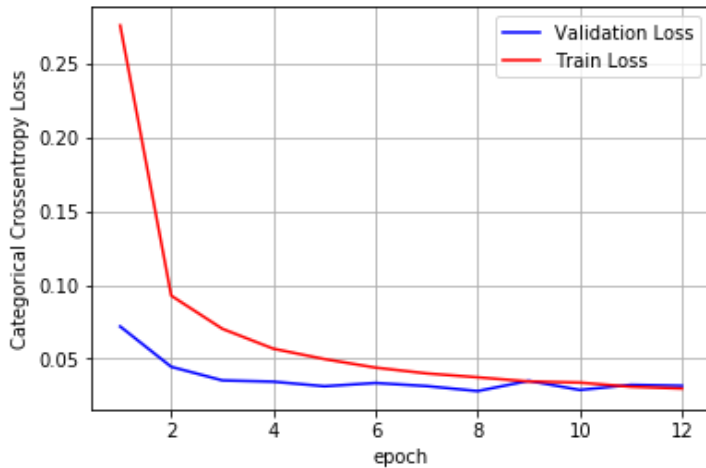
```

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')
x = list(range(1, epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)

```

Test score: 0.03149389391231671

Test accuracy: 0.9897

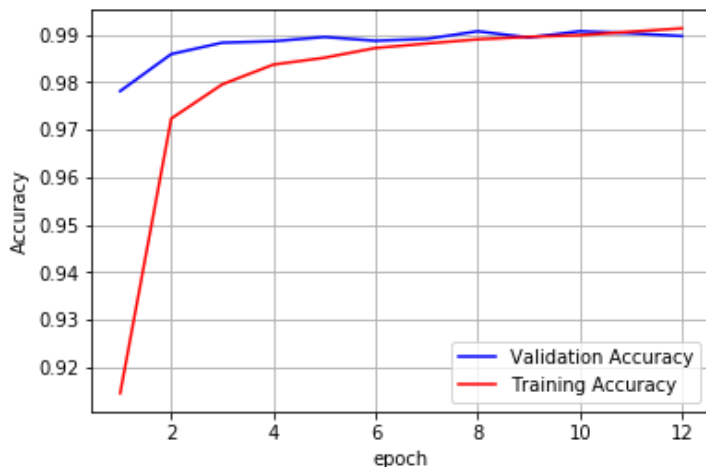


In [0]:

```

fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1, epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)

```



2. 2 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION)

In [0]:

```

from keras.layers.normalization import BatchNormalization
model2 = Sequential()

model2.add(Conv2D(64, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape))
model2.add(BatchNormalization())
model2.add(MaxPooling2D(pool_size=(2, 2)))

model2.add(Conv2D(32, (3, 3), activation='relu'))
model2.add(BatchNormalization())
model2.add(MaxPooling2D(pool_size=(2, 2)))

```

```

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

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

inspect = model2.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    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
60000/60000 [=====] - 5s 90us/step - loss: 0.2421 - acc: 0.9275
- val_loss: 0.0588 - val_acc: 0.9808
Epoch 2/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0849 - acc: 0.9748
- val_loss: 0.0432 - val_acc: 0.9862
Epoch 3/12
60000/60000 [=====] - 4s 74us/step - loss: 0.0603 - acc: 0.9819
- val_loss: 0.0310 - val_acc: 0.9898
Epoch 4/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0499 - acc: 0.9846
- val_loss: 0.0492 - val_acc: 0.9858
Epoch 5/12
60000/60000 [=====] - 4s 74us/step - loss: 0.0414 - acc: 0.9872
- val_loss: 0.0360 - val_acc: 0.9889
Epoch 6/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0381 - acc: 0.9882
- val_loss: 0.0285 - val_acc: 0.9913
Epoch 7/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0323 - acc: 0.9900
- val_loss: 0.0355 - val_acc: 0.9910
Epoch 8/12
60000/60000 [=====] - 4s 74us/step - loss: 0.0283 - acc: 0.9914
- val_loss: 0.0509 - val_acc: 0.9873
Epoch 9/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0275 - acc: 0.9919
- val_loss: 0.0264 - val_acc: 0.9922
Epoch 10/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0248 - acc: 0.9923
- val_loss: 0.0256 - val_acc: 0.9922
Epoch 11/12
60000/60000 [=====] - 4s 75us/step - loss: 0.0215 - acc: 0.9933
- val_loss: 0.0245 - val_acc: 0.9923
Epoch 12/12
60000/60000 [=====] - 4s 74us/step - loss: 0.0200 - acc: 0.9938
- val_loss: 0.0235 - val_acc: 0.9936
Test loss: 0.023466705693548782
Test accuracy: 0.9936

```

In [0]:

```
model2.summary()
```

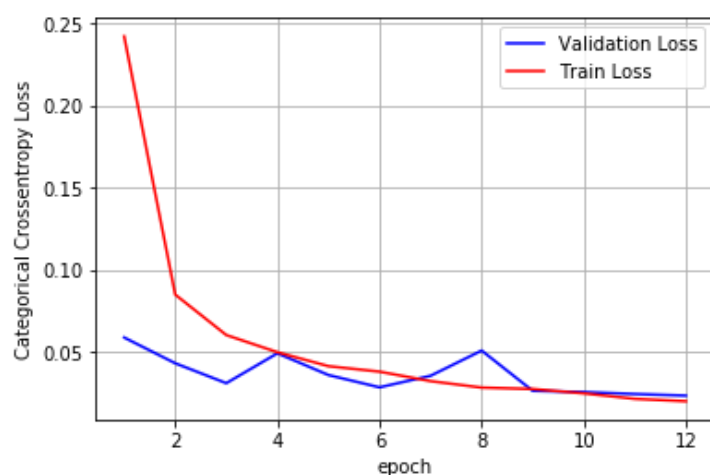
| Layer (type) | Output Shape | Param # |
|---|--------------------|---------|
| conv2d_11 (Conv2D) | (None, 26, 26, 64) | 640 |
| batch_normalization_3 (Batch Normalization) | (None, 26, 26, 64) | 256 |
| max_pooling2d_7 (MaxPooling2D) | (None, 13, 13, 64) | 0 |

| | | |
|---|--------------------|--------|
| conv2d_12 (Conv2D) | (None, 11, 11, 32) | 18464 |
| batch_normalization_4 (Batch Normalization) | (None, 11, 11, 32) | 128 |
| max_pooling2d_8 (MaxPooling2D) | (None, 5, 5, 32) | 0 |
| flatten_6 (Flatten) | (None, 800) | 0 |
| dense_11 (Dense) | (None, 128) | 102528 |
| dropout_10 (Dropout) | (None, 128) | 0 |
| dense_12 (Dense) | (None, 10) | 1290 |
| ===== | | |
| Total params: 123,306 | | |
| Trainable params: 123,114 | | |
| Non-trainable params: 192 | | |

In [0]:

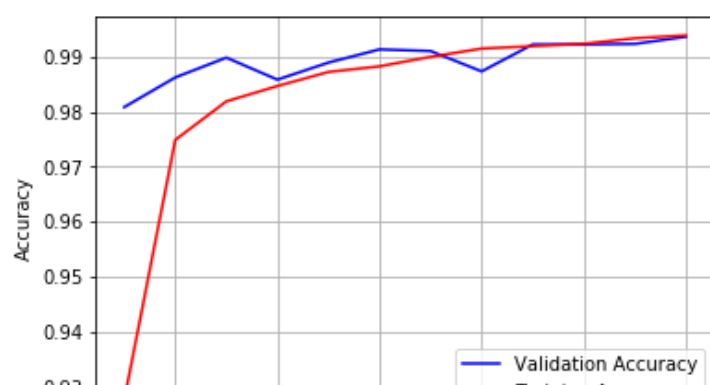
```
score = model2.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

Test score: 0.023466705693548782
Test accuracy: 0.9936



In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```





3. 2 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION AND DROPOUT)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model3 = Sequential()

model3.add(Conv2D(64, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape))
model3.add(BatchNormalization())
model3.add(MaxPooling2D(pool_size=(2, 2)))
model3.add(Dropout(0.5))

model3.add(Conv2D(32, (3, 3), activation='relu'))
model3.add(BatchNormalization())
model3.add(MaxPooling2D(pool_size=(2, 2)))
model3.add(Dropout(0.5))

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

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

inspect = model3.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    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 1/12
60000/60000 [=====] - 6s 100us/step - loss: 0.6806 - acc: 0.7886
- val_loss: 0.1084 - val_acc: 0.9656
Epoch 2/12
60000/60000 [=====] - 5s 81us/step - loss: 0.2121 - acc: 0.9370
- val_loss: 0.0637 - val_acc: 0.9802
Epoch 3/12
60000/60000 [=====] - 5s 81us/step - loss: 0.1523 - acc: 0.9553
- val_loss: 0.0481 - val_acc: 0.9837
Epoch 4/12
60000/60000 [=====] - 5s 82us/step - loss: 0.1286 - acc: 0.9618
- val_loss: 0.0389 - val_acc: 0.9865
Epoch 5/12
60000/60000 [=====] - 5s 81us/step - loss: 0.1175 - acc: 0.9653
- val_loss: 0.0375 - val_acc: 0.9886
Epoch 6/12
60000/60000 [=====] - 5s 81us/step - loss: 0.1036 - acc: 0.9696
- val_loss: 0.0375 - val_acc: 0.9873
Epoch 7/12
60000/60000 [=====] - 5s 82us/step - loss: 0.1018 - acc: 0.9708
- val_loss: 0.0377 - val_acc: 0.9887
Epoch 8/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0955 - acc: 0.9722
- val_loss: 0.0374 - val_acc: 0.9886
Epoch 9/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0871 - acc: 0.9745
- val_loss: 0.0320 - val_acc: 0.9900
Epoch 10/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0843 - acc: 0.9747
```

```

- val_loss: 0.0288 - val_acc: 0.9903
Epoch 11/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0805 - acc: 0.9761
- val_loss: 0.0328 - val_acc: 0.9884
Epoch 12/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0778 - acc: 0.9771
- val_loss: 0.0323 - val_acc: 0.9896
Test loss: 0.03226553132739282
Test accuracy: 0.9896

```

In [0]:

```
model3.summary()
```

| Layer (type) | Output Shape | Param # |
|---|--------------------|---------|
| conv2d_13 (Conv2D) | (None, 26, 26, 64) | 640 |
| batch_normalization_5 (Batch Normalization) | (None, 26, 26, 64) | 256 |
| max_pooling2d_9 (MaxPooling2D) | (None, 13, 13, 64) | 0 |
| dropout_11 (Dropout) | (None, 13, 13, 64) | 0 |
| conv2d_14 (Conv2D) | (None, 11, 11, 32) | 18464 |
| batch_normalization_6 (Batch Normalization) | (None, 11, 11, 32) | 128 |
| max_pooling2d_10 (MaxPooling2D) | (None, 5, 5, 32) | 0 |
| dropout_12 (Dropout) | (None, 5, 5, 32) | 0 |
| flatten_7 (Flatten) | (None, 800) | 0 |
| dense_13 (Dense) | (None, 128) | 102528 |
| dropout_13 (Dropout) | (None, 128) | 0 |
| dense_14 (Dense) | (None, 10) | 1290 |
| Total params: 123,306 | | |
| Trainable params: 123,114 | | |
| Non-trainable params: 192 | | |

In [0]:

```

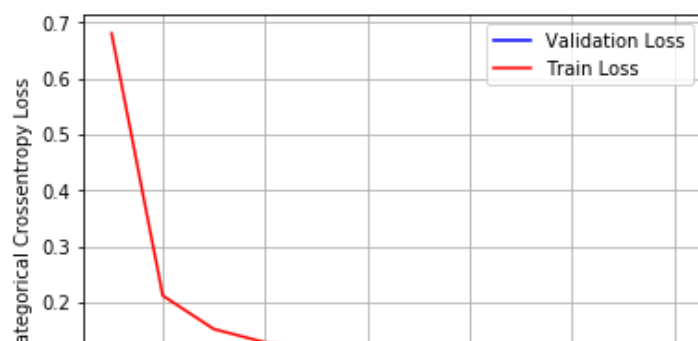
score = model3.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1, epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)

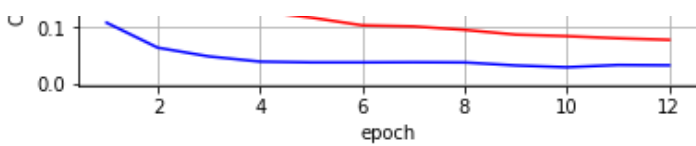
```

```

Test score: 0.03226553132739282
Test accuracy: 0.9896

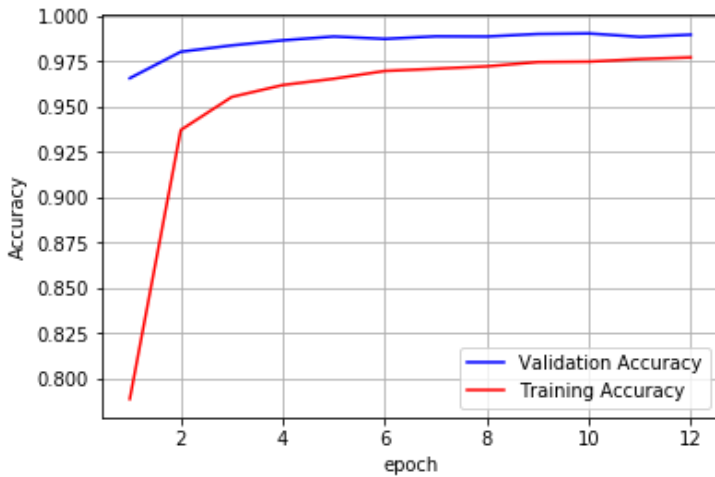
```





In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



4. 3 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION 2X2 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model4 = Sequential()

model4.add(Conv2D(64, kernel_size=(2, 2),
                  activation='relu',
                  input_shape=input_shape))
model4.add(BatchNormalization())
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Conv2D(32, (2, 2), activation='relu'))
model4.add(BatchNormalization())
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Conv2D(32, (2, 2), activation='relu'))
model4.add(BatchNormalization())
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Flatten())
model4.add(Dense(128, activation='relu'))
model4.add(Dropout(0.5))
model4.add(Dense(num_classes, activation='softmax'))

model4.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model4.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model4.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 [=====] - 7s 108us/step - loss: 0.3744 - acc: 0.8829
- val_loss: 0.1285 - val_acc: 0.9619
Epoch 2/12
60000/60000 [=====] - 5s 81us/step - loss: 0.1192 - acc: 0.9644
- val_loss: 0.0635 - val_acc: 0.9817
Epoch 3/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0925 - acc: 0.9725
- val_loss: 0.0627 - val_acc: 0.9794
Epoch 4/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0728 - acc: 0.9781
- val_loss: 0.0570 - val_acc: 0.9840
Epoch 5/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0648 - acc: 0.9810
- val_loss: 0.0456 - val_acc: 0.9858
Epoch 6/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0559 - acc: 0.9830
- val_loss: 0.0432 - val_acc: 0.9867
Epoch 7/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0508 - acc: 0.9846
- val_loss: 0.0452 - val_acc: 0.9865
Epoch 8/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0474 - acc: 0.9856
- val_loss: 0.0487 - val_acc: 0.9864
Epoch 9/12
60000/60000 [=====] - 5s 81us/step - loss: 0.0422 - acc: 0.9867
- val_loss: 0.0476 - val_acc: 0.9864
Epoch 10/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0422 - acc: 0.9872
- val_loss: 0.0471 - val_acc: 0.9857
Epoch 11/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0376 - acc: 0.9890
- val_loss: 0.0394 - val_acc: 0.9889
Epoch 12/12
60000/60000 [=====] - 5s 80us/step - loss: 0.0342 - acc: 0.9895
- val_loss: 0.0450 - val_acc: 0.9874
Test loss: 0.045036793098763704
Test accuracy: 0.9874

In [0]:

```
model4.summary()
```

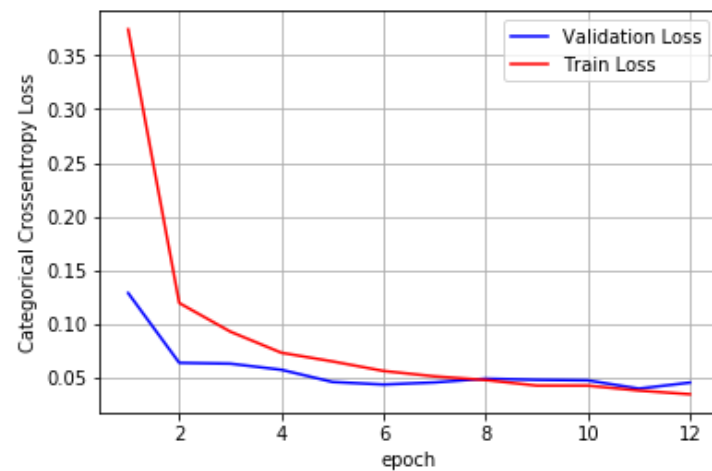
| Layer (type) | Output Shape | Param # |
|--|--------------------|---------|
| conv2d_21 (Conv2D) | (None, 27, 27, 64) | 320 |
| batch_normalization_13 (Batch Normalization) | (None, 27, 27, 64) | 256 |
| max_pooling2d_17 (MaxPooling2D) | (None, 13, 13, 64) | 0 |
| conv2d_22 (Conv2D) | (None, 12, 12, 32) | 8224 |
| batch_normalization_14 (Batch Normalization) | (None, 12, 12, 32) | 128 |
| max_pooling2d_18 (MaxPooling2D) | (None, 6, 6, 32) | 0 |
| conv2d_23 (Conv2D) | (None, 5, 5, 32) | 4128 |
| batch_normalization_15 (Batch Normalization) | (None, 5, 5, 32) | 128 |
| max_pooling2d_19 (MaxPooling2D) | (None, 2, 2, 32) | 0 |
| flatten_10 (Flatten) | (None, 128) | 0 |
| dense_19 (Dense) | (None, 128) | 16512 |
| dropout_19 (Dropout) | (None, 128) | 0 |
| dense_20 (Dense) | (None, 10) | 1290 |

Total params: 30,986
Trainable params: 30,730
Non-trainable params: 256

In [0]:

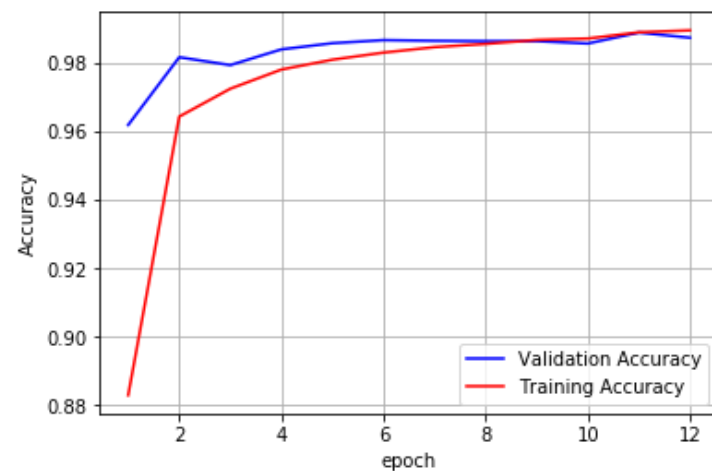
```
score = model4.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)
```

Test score: 0.045036793098763704
Test accuracy: 0.9874



In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



5. 3 CONVOLUTIONAL LAYERS (MAX POOLING,DROPOUT AND BATCH NORMALIZATION 5X5 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model5 = Sequential()
#1ST
```

```

model5.add(Conv2D(128, kernel_size=(5, 5),
                  activation='relu',
                  input_shape=input_shape))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
model5.add(Dropout(0.5))

#2ND
model5.add(Conv2D(64, (5, 5), activation='relu'))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
model5.add(Dropout(0.5))

#3RD

model5.add(Conv2D(32, (3, 3), activation='relu'))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
model5.add(Dropout(0.5))

model5.add(Flatten())
model5.add(Dense(128, activation='relu'))
model5.add(Dropout(0.5))
model5.add(Dense(num_classes, activation='softmax'))

model5.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model5.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model5.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 [=====] - 9s 151us/step - loss: 0.8208 - acc: 0.7304
- val_loss: 0.0687 - val_acc: 0.9793
Epoch 2/12
60000/60000 [=====] - 7s 118us/step - loss: 0.2279 - acc: 0.9324
- val_loss: 0.0879 - val_acc: 0.9726
Epoch 3/12
60000/60000 [=====] - 7s 119us/step - loss: 0.1650 - acc: 0.9520
- val_loss: 0.0400 - val_acc: 0.9878
Epoch 4/12
60000/60000 [=====] - 7s 118us/step - loss: 0.1426 - acc: 0.9598
- val_loss: 0.0365 - val_acc: 0.9892
Epoch 5/12
60000/60000 [=====] - 7s 119us/step - loss: 0.1218 - acc: 0.9652
- val_loss: 0.0289 - val_acc: 0.9903
Epoch 6/12
60000/60000 [=====] - 7s 119us/step - loss: 0.1123 - acc: 0.9684
- val_loss: 0.0273 - val_acc: 0.9915
Epoch 7/12
60000/60000 [=====] - 7s 118us/step - loss: 0.1044 - acc: 0.9715
- val_loss: 0.0275 - val_acc: 0.9918
Epoch 8/12
60000/60000 [=====] - 7s 119us/step - loss: 0.0985 - acc: 0.9728
- val_loss: 0.0262 - val_acc: 0.9909
Epoch 9/12
60000/60000 [=====] - 7s 118us/step - loss: 0.0961 - acc: 0.9734
- val_loss: 0.0236 - val_acc: 0.9935
Epoch 10/12
60000/60000 [=====] - 7s 118us/step - loss: 0.0901 - acc: 0.9751
- val_loss: 0.0287 - val_acc: 0.9911
Epoch 11/12
60000/60000 [=====] - 7s 118us/step - loss: 0.0831 - acc: 0.9764
- val_loss: 0.0231 - val_acc: 0.9924

```

Epoch 12/12
60000/60000 [=====] - 7s 118us/step - loss: 0.0846 - acc: 0.9765
- val_loss: 0.0235 - val_acc: 0.9927
Test loss: 0.023486822005478462
Test accuracy: 0.9927

In [0]:

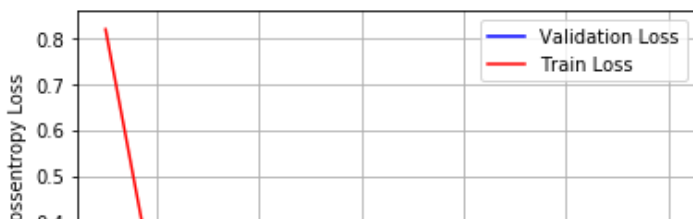
```
model5.summary()
```

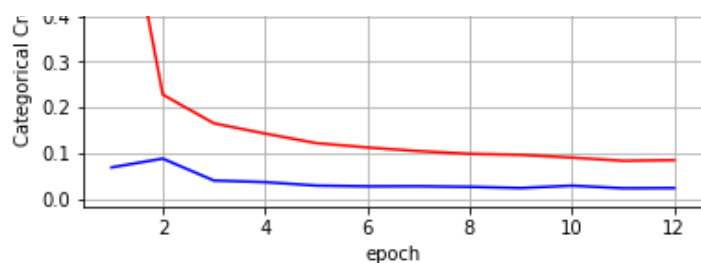
| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| conv2d_24 (Conv2D) | (None, 24, 24, 128) | 3328 |
| batch_normalization_16 (Batch Normalization) | (None, 24, 24, 128) | 512 |
| max_pooling2d_20 (MaxPooling2D) | (None, 12, 12, 128) | 0 |
| dropout_20 (Dropout) | (None, 12, 12, 128) | 0 |
| conv2d_25 (Conv2D) | (None, 8, 8, 64) | 204864 |
| batch_normalization_17 (Batch Normalization) | (None, 8, 8, 64) | 256 |
| max_pooling2d_21 (MaxPooling2D) | (None, 4, 4, 64) | 0 |
| dropout_21 (Dropout) | (None, 4, 4, 64) | 0 |
| conv2d_26 (Conv2D) | (None, 2, 2, 32) | 18464 |
| batch_normalization_18 (Batch Normalization) | (None, 2, 2, 32) | 128 |
| max_pooling2d_22 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| dropout_22 (Dropout) | (None, 1, 1, 32) | 0 |
| flatten_11 (Flatten) | (None, 32) | 0 |
| dense_21 (Dense) | (None, 128) | 4224 |
| dropout_23 (Dropout) | (None, 128) | 0 |
| dense_22 (Dense) | (None, 10) | 1290 |
| Total params: 233,066 | | |
| Trainable params: 232,618 | | |
| Non-trainable params: 448 | | |

In [0]:

```
score = model5.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')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

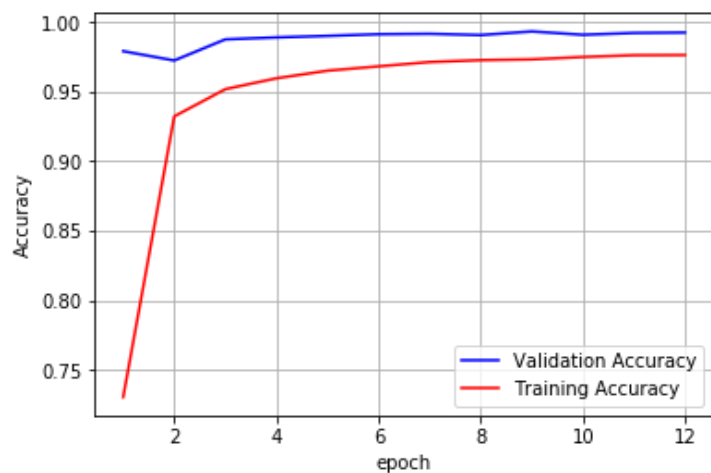
Test score: 0.023486822005478462
Test accuracy: 0.9927





In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



6. 3 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION 7X7 AND 5X5 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model6 = Sequential()
#1ST
model6.add(Conv2D(128, kernel_size=(7, 7),
                  activation='relu',
                  input_shape=input_shape))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))

#2ND
model6.add(Conv2D(64, (5, 5), activation='relu'))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))

#3RD
model6.add(Conv2D(32, (3, 3), activation='relu'))
model6.add(BatchNormalization())
#model.add(MaxPooling2D(pool_size=(2, 2)))

model6.add(Flatten())
model6.add(Dense(128, activation='relu'))
model6.add(Dense(64, activation='relu'))
model6.add(Dense(num_classes, activation='softmax'))

model6.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])
```

```
inspect = model6.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model6.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 [=====] - 9s 146us/step - loss: 0.1286 - acc: 0.9657
- val_loss: 0.0610 - val_acc: 0.9820
Epoch 2/12
60000/60000 [=====] - 6s 105us/step - loss: 0.0386 - acc: 0.9878
- val_loss: 0.0315 - val_acc: 0.9899
Epoch 3/12
60000/60000 [=====] - 6s 107us/step - loss: 0.0262 - acc: 0.9913
- val_loss: 0.0456 - val_acc: 0.9879
Epoch 4/12
60000/60000 [=====] - 6s 104us/step - loss: 0.0193 - acc: 0.9937
- val_loss: 0.0295 - val_acc: 0.9907
Epoch 5/12
60000/60000 [=====] - 6s 105us/step - loss: 0.0132 - acc: 0.9957
- val_loss: 0.0390 - val_acc: 0.9887
Epoch 6/12
60000/60000 [=====] - 6s 105us/step - loss: 0.0101 - acc: 0.9970
- val_loss: 0.0271 - val_acc: 0.9916
Epoch 7/12
60000/60000 [=====] - 6s 105us/step - loss: 0.0077 - acc: 0.9975
- val_loss: 0.0253 - val_acc: 0.9928
Epoch 8/12
60000/60000 [=====] - 6s 104us/step - loss: 0.0058 - acc: 0.9980
- val_loss: 0.0354 - val_acc: 0.9912
Epoch 9/12
60000/60000 [=====] - 6s 105us/step - loss: 0.0049 - acc: 0.9984
- val_loss: 0.0295 - val_acc: 0.9923
Epoch 10/12
60000/60000 [=====] - 6s 104us/step - loss: 0.0033 - acc: 0.9987
- val_loss: 0.0359 - val_acc: 0.9920
Epoch 11/12
60000/60000 [=====] - 6s 104us/step - loss: 0.0033 - acc: 0.9989
- val_loss: 0.0303 - val_acc: 0.9931
Epoch 12/12
60000/60000 [=====] - 6s 104us/step - loss: 0.0027 - acc: 0.9990
- val_loss: 0.0332 - val_acc: 0.9927
Test loss: 0.03316479856713909
Test accuracy: 0.9927
```

In [0]:

```
model6.summary()
```

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_33 (Conv2D) | (None, 22, 22, 128) | 6400 |
| batch_normalization_25 (Batch Normalization) | (None, 22, 22, 128) | 512 |
| max_pooling2d_27 (MaxPooling2D) | (None, 11, 11, 128) | 0 |
| conv2d_34 (Conv2D) | (None, 7, 7, 64) | 204864 |
| batch_normalization_26 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_28 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| conv2d_35 (Conv2D) | (None, 1, 1, 32) | 18464 |
| batch_normalization_27 (Batch Normalization) | (None, 1, 1, 32) | 128 |

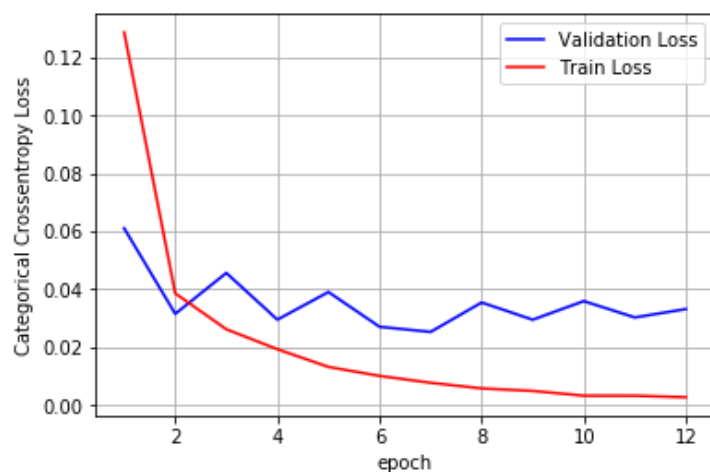
| | | |
|---------------------------|-------------|------|
| flatten_8 (Flatten) | (None, 32) | 0 |
| dense_15 (Dense) | (None, 128) | 4224 |
| dense_16 (Dense) | (None, 64) | 8256 |
| dense_17 (Dense) | (None, 10) | 650 |
| ===== | | |
| Total params: 243,754 | | |
| Trainable params: 243,306 | | |
| Non-trainable params: 448 | | |

In [0]:

```
score = model6.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')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

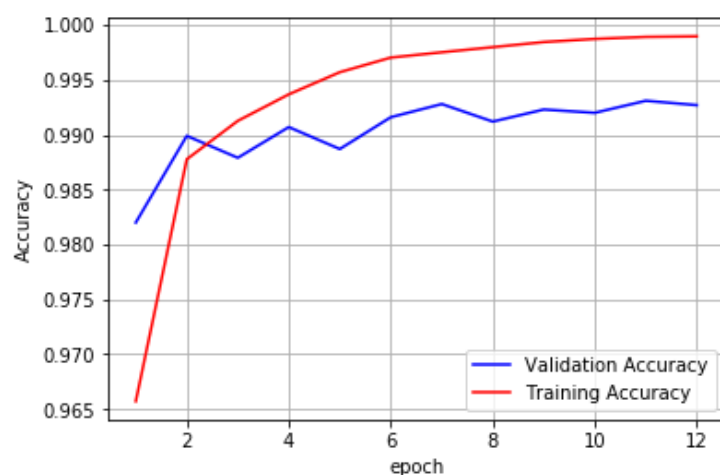
Test score: 0.03316479856713909

Test accuracy: 0.9927



In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



7. 5 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION 3X3 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model7 = Sequential()
#1ST
model7.add(Conv2D(256, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape,padding="same"))
model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))

#2ND
model7.add(Conv2D(128, (3, 3), activation='relu',padding="same"))
model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))

#3RD
model7.add(Conv2D(64, (3, 3), activation='relu',padding="same"))
model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))

#4TH
model7.add(Conv2D(32, (3, 3), activation='relu',padding="same"))
model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))

#5TH
model7.add(Conv2D(16, (3, 3), activation='relu',padding="same"))
model7.add(BatchNormalization())
#model.add(MaxPooling2D(pool_size=(2, 2)))

model7.add(Flatten())
model7.add(Dense(32, activation='relu'))

model7.add(Dense(num_classes, activation='softmax'))

model7.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model7.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model7.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 [=====] - 18s 294us/step - loss: 0.2058 - acc: 0.951
2 - val_loss: 0.0864 - val_acc: 0.9748

Epoch 2/12

60000/60000 [=====] - 15s 247us/step - loss: 0.0435 - acc: 0.987
3 - val_loss: 0.0550 - val_acc: 0.9841

Epoch 3/12

60000/60000 [=====] - 15s 247us/step - loss: 0.0289 - acc: 0.991
5 - val_loss: 0.0323 - val_acc: 0.9903

Epoch 4/12

60000/60000 [=====] - 15s 247us/step - loss: 0.0202 - acc: 0.993
7 - val_loss: 0.0297 - val_acc: 0.9907

Epoch 5/12

60000/60000 [=====] - 15s 246us/step - loss: 0.0139 - acc: 0.996
0 - val_loss: 0.1061 - val_acc: 0.9695

Epoch 6/12

60000/60000 [=====] - 15s 246us/step - loss: 0.0113 - acc: 0.996
5 - val_loss: 0.0327 - val_acc: 0.9908

```

Epoch 7/12
60000/60000 [=====] - 15s 246us/step - loss: 0.0079 - acc: 0.997
7 - val_loss: 0.0266 - val_acc: 0.9933
Epoch 8/12
60000/60000 [=====] - 15s 247us/step - loss: 0.0060 - acc: 0.998
3 - val_loss: 0.0270 - val_acc: 0.9923
Epoch 9/12
60000/60000 [=====] - 15s 246us/step - loss: 0.0057 - acc: 0.998
0 - val_loss: 0.0377 - val_acc: 0.9900
Epoch 10/12
60000/60000 [=====] - 15s 246us/step - loss: 0.0031 - acc: 0.999
2 - val_loss: 0.0342 - val_acc: 0.9906
Epoch 11/12
60000/60000 [=====] - 15s 246us/step - loss: 0.0027 - acc: 0.999
3 - val_loss: 0.0361 - val_acc: 0.9908
Epoch 12/12
60000/60000 [=====] - 15s 246us/step - loss: 0.0016 - acc: 0.999
6 - val_loss: 0.0321 - val_acc: 0.9913
Test loss: 0.032069397589693835
Test accuracy: 0.9913

```

In [0]:

```
model7.summary()
```

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_30 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_22 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_25 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| conv2d_31 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_23 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_26 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| conv2d_32 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_24 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_27 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| conv2d_33 (Conv2D) | (None, 3, 3, 32) | 18464 |
| batch_normalization_25 (Batch Normalization) | (None, 3, 3, 32) | 128 |
| max_pooling2d_28 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| conv2d_34 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_26 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| flatten_13 (Flatten) | (None, 16) | 0 |
| dense_26 (Dense) | (None, 32) | 544 |
| dense_27 (Dense) | (None, 10) | 330 |
| ===== | | |
| Total params: 397,338 | | |
| Trainable params: 396,346 | | |
| Non-trainable params: 992 | | |

In [0]:

```

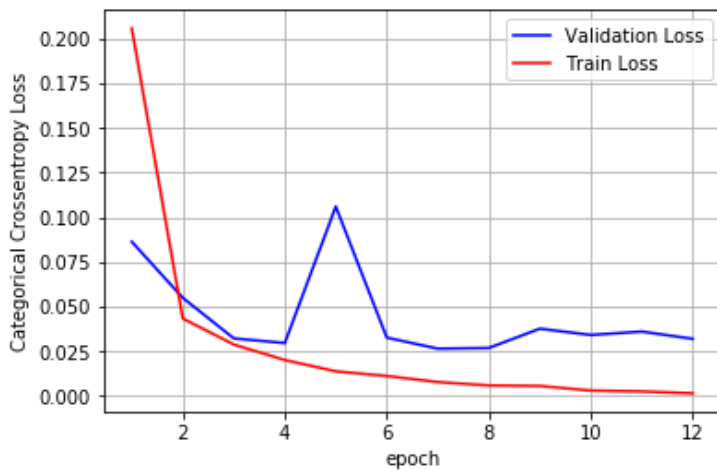
score = model7.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])

```

```
print('Test accuracy:', score[1])
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

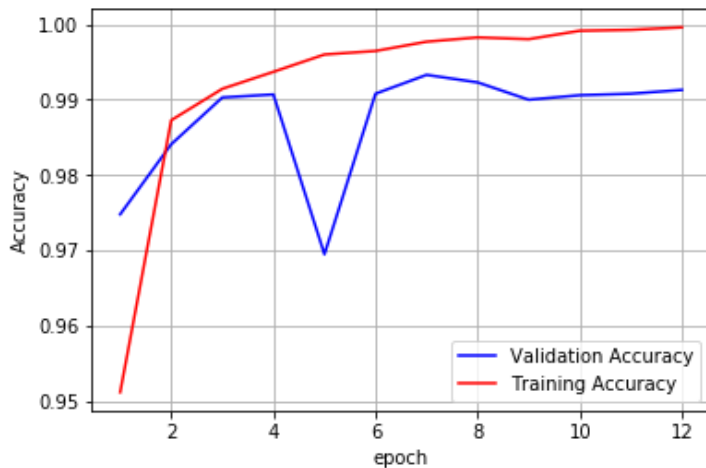
Test score: 0.032069397589693835

Test accuracy: 0.9913



In [0]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



8. 5 CONVOLUTIONAL LAYERS (MAX POOLING,DROPOUT AND BATCH NORMALIZATION 3X3 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model8 = Sequential()
#1ST
model8.add(Conv2D(256, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape,padding="same"))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#2ND
model8.add(Conv2D(128, (3, 3), activation='relu',padding="same"))
```

```

model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#3RD
model8.add(Conv2D(64, (3, 3), activation='relu',padding="same"))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#4TH
model8.add(Conv2D(32, (3, 3), activation='relu',padding="same"))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#5TH
model8.add(Conv2D(16, (3, 3), activation='relu',padding="same"))
model8.add(BatchNormalization())
#model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

model8.add(Flatten())
model8.add(Dense(32, activation='relu'))

model8.add(Dense(num_classes, activation='softmax'))

model8.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model8.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model8.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 [=====] - 19s 321us/step - loss: 1.5369 - acc: 0.440
7 - val_loss: 0.4345 - val_acc: 0.9056
Epoch 2/12
60000/60000 [=====] - 17s 275us/step - loss: 0.6787 - acc: 0.756
6 - val_loss: 0.1479 - val_acc: 0.9698
Epoch 3/12
60000/60000 [=====] - 16s 275us/step - loss: 0.4721 - acc: 0.837
4 - val_loss: 0.1065 - val_acc: 0.9716
Epoch 4/12
60000/60000 [=====] - 16s 274us/step - loss: 0.3915 - acc: 0.869
5 - val_loss: 0.0564 - val_acc: 0.9841
Epoch 5/12
60000/60000 [=====] - 16s 275us/step - loss: 0.3410 - acc: 0.886
6 - val_loss: 0.0594 - val_acc: 0.9841
Epoch 6/12
60000/60000 [=====] - 16s 275us/step - loss: 0.3148 - acc: 0.894
9 - val_loss: 0.0508 - val_acc: 0.9868
Epoch 7/12
60000/60000 [=====] - 16s 274us/step - loss: 0.2894 - acc: 0.901
2 - val_loss: 0.0454 - val_acc: 0.9888
Epoch 8/12
60000/60000 [=====] - 16s 274us/step - loss: 0.2782 - acc: 0.905
6 - val_loss: 0.0442 - val_acc: 0.9890
Epoch 9/12
60000/60000 [=====] - 16s 273us/step - loss: 0.2633 - acc: 0.912
4 - val_loss: 0.0436 - val_acc: 0.9891
Epoch 10/12
60000/60000 [=====] - 17s 276us/step - loss: 0.2566 - acc: 0.914
4 - val_loss: 0.0411 - val_acc: 0.9896

```

```
Epoch 11/12
60000/60000 [=====] - 16s 275us/step - loss: 0.2422 - acc: 0.919
0 - val_loss: 0.0385 - val_acc: 0.9896
Epoch 12/12
60000/60000 [=====] - 16s 273us/step - loss: 0.2344 - acc: 0.920
5 - val_loss: 0.0385 - val_acc: 0.9900
Test loss: 0.0385118981421052
Test accuracy: 0.99
```

In [0]:

```
model8.summary()
```

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_35 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_27 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_29 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| dropout_24 (Dropout) | (None, 14, 14, 256) | 0 |
| conv2d_36 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_28 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_30 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| dropout_25 (Dropout) | (None, 7, 7, 128) | 0 |
| conv2d_37 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_29 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_31 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| dropout_26 (Dropout) | (None, 3, 3, 64) | 0 |
| conv2d_38 (Conv2D) | (None, 3, 3, 32) | 18464 |
| batch_normalization_30 (Batch Normalization) | (None, 3, 3, 32) | 128 |
| max_pooling2d_32 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| dropout_27 (Dropout) | (None, 1, 1, 32) | 0 |
| conv2d_39 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_31 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| dropout_28 (Dropout) | (None, 1, 1, 16) | 0 |
| flatten_14 (Flatten) | (None, 16) | 0 |
| dense_28 (Dense) | (None, 32) | 544 |
| dense_29 (Dense) | (None, 10) | 330 |
| ===== | | |
| Total params: 397,338 | | |
| Trainable params: 396,346 | | |
| Non-trainable params: 992 | | |

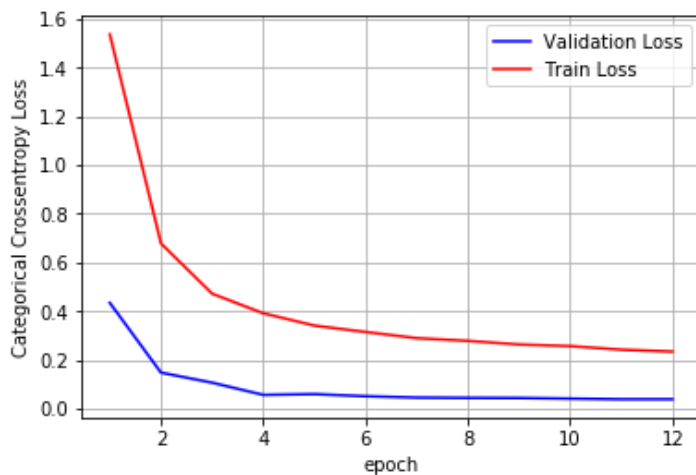
In [0]:

```
score = model8.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')
```

```
x = list(range(1, epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

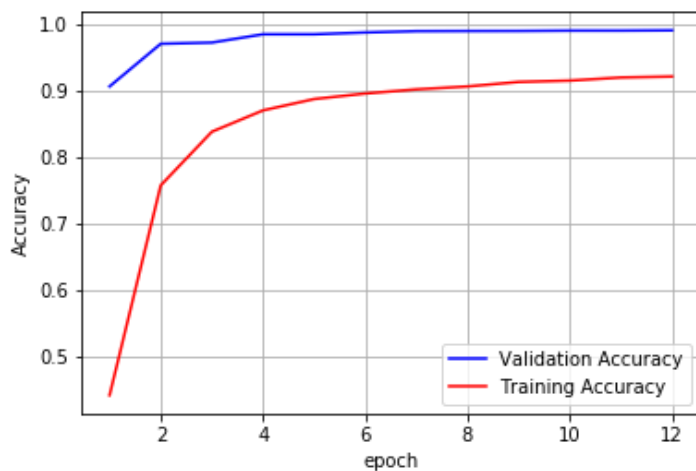
Test score: 0.0385118981421052

Test accuracy: 0.99



In [0]:

```
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1, epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



9. 7 CONVOLUTIONAL LAYERS (MAX POOLING,DROPOUT AND BATCH NORMALIZATION 3X3 FILTER)

In [0]:

```
from keras.layers.normalization import BatchNormalization
model9 = Sequential()
#1ST
model9.add(Conv2D(256, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape, padding="same"))
model9.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

#2ND
model9.add(Conv2D(128, (3, 3), activation='relu', padding="same"))
model9.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))
```

```

#3RD
model9.add(Conv2D(64, (3, 3), activation='relu',padding="same"))
model9.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

#4TH
model9.add(Conv2D(64, (3, 3), activation='relu',padding="same"))
model9.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

#5TH
model9.add(Conv2D(32, (3, 3), activation='relu',padding="same"))
model9.add(BatchNormalization())
#model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

#6TH
model9.add(Conv2D(32, (3, 3), activation='relu',padding="same"))
model9.add(BatchNormalization())
#model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

#7TH
model9.add(Conv2D(16, (3, 3), activation='relu',padding="same"))
model9.add(BatchNormalization())
#model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.2))

model9.add(Flatten())
model9.add(Dense(32, activation='relu'))

model9.add(Dense(num_classes, activation='softmax'))

model9.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model9.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model9.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 [=====] - 21s 347us/step - loss: 0.6060 - acc: 0.8114 - val_loss: 0.0880 - val_acc: 0.9770

Epoch 2/12

60000/60000 [=====] - 17s 288us/step - loss: 0.1467 - acc: 0.9634 - val_loss: 0.0659 - val_acc: 0.9816

Epoch 3/12

60000/60000 [=====] - 18s 295us/step - loss: 0.1015 - acc: 0.9753 - val_loss: 0.0462 - val_acc: 0.9891

Epoch 4/12

60000/60000 [=====] - 18s 296us/step - loss: 0.0836 - acc: 0.9798 - val_loss: 0.0354 - val_acc: 0.9917

Epoch 5/12

60000/60000 [=====] - 18s 297us/step - loss: 0.0704 - acc: 0.9830 - val_loss: 0.0304 - val_acc: 0.9923

Epoch 6/12

60000/60000 [=====] - 18s 296us/step - loss: 0.0639 - acc: 0.9851 - val_loss: 0.0315 - val_acc: 0.9921

Epoch 7/12

60000/60000 [=====] - 17s 290us/step - loss: 0.0568 - acc: 0.9866 - val_loss: 0.0290 - val_acc: 0.9930

```

Epoch 8/12
60000/60000 [=====] - 17s 291us/step - loss: 0.0481 - acc: 0.988
3 - val_loss: 0.0373 - val_acc: 0.9914
Epoch 9/12
60000/60000 [=====] - 17s 292us/step - loss: 0.0488 - acc: 0.988
4 - val_loss: 0.0274 - val_acc: 0.9938
Epoch 10/12
60000/60000 [=====] - 18s 296us/step - loss: 0.0449 - acc: 0.989
3 - val_loss: 0.0256 - val_acc: 0.9932
Epoch 11/12
60000/60000 [=====] - 18s 292us/step - loss: 0.0419 - acc: 0.990
5 - val_loss: 0.0454 - val_acc: 0.9909
Epoch 12/12
60000/60000 [=====] - 17s 291us/step - loss: 0.0388 - acc: 0.990
4 - val_loss: 0.0254 - val_acc: 0.9938
Test loss: 0.025403328641527334
Test accuracy: 0.9938

```

In [0]:

```
model9.summary()
```

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_40 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_32 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_33 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| dropout_29 (Dropout) | (None, 14, 14, 256) | 0 |
| conv2d_41 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_33 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_34 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| dropout_30 (Dropout) | (None, 7, 7, 128) | 0 |
| conv2d_42 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_34 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_35 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| dropout_31 (Dropout) | (None, 3, 3, 64) | 0 |
| conv2d_43 (Conv2D) | (None, 3, 3, 64) | 36928 |
| batch_normalization_35 (Batch Normalization) | (None, 3, 3, 64) | 256 |
| max_pooling2d_36 (MaxPooling2D) | (None, 1, 1, 64) | 0 |
| dropout_32 (Dropout) | (None, 1, 1, 64) | 0 |
| conv2d_44 (Conv2D) | (None, 1, 1, 32) | 18464 |
| batch_normalization_36 (Batch Normalization) | (None, 1, 1, 32) | 128 |
| dropout_33 (Dropout) | (None, 1, 1, 32) | 0 |
| conv2d_45 (Conv2D) | (None, 1, 1, 32) | 9248 |
| batch_normalization_37 (Batch Normalization) | (None, 1, 1, 32) | 128 |
| dropout_34 (Dropout) | (None, 1, 1, 32) | 0 |
| conv2d_46 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_38 (Batch Normalization) | (None, 1, 1, 16) | 64 |

| | | |
|-----------------------------|------------------|-----|
| dropout_35 (Dropout) | (None, 1, 1, 16) | 0 |
| flatten_15 (Flatten) | (None, 16) | 0 |
| dense_30 (Dense) | (None, 32) | 544 |
| dense_31 (Dense) | (None, 10) | 330 |
| ===== | | |
| Total params: 443,898 | | |
| Trainable params: 442,714 | | |
| Non-trainable params: 1,184 | | |

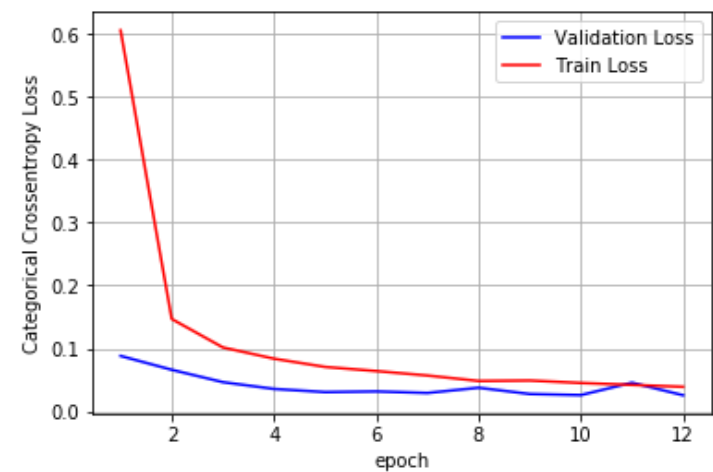
In [0]:

```

score = model9.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')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)

```

Test score: 0.025403328641527334
Test accuracy: 0.9938

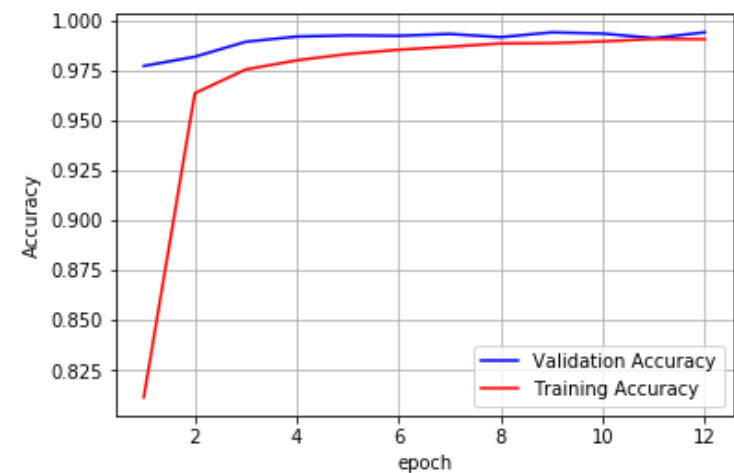


In [0]:

```

fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)

```



```
In [0]:
```

```
x= [model1.evaluate(x_test, y_test, verbose=0)[0],model2.evaluate(x_test, y_test, verbose=0)[0],model3.evaluate(x_test, y_test, verbose=0)[0],model4.evaluate(x_test, y_test, verbose=0)[0],model5.evaluate(x_test, y_test, verbose=0)[0],model6.evaluate(x_test, y_test, verbose=0)[0],model7.evaluate(x_test, y_test, verbose=0)[0],model8.evaluate(x_test, y_test, verbose=0)[0],model9.evaluate(x_test, y_test, verbose=0)[0]]
```

```
In [0]:
```

```
y= [model1.evaluate(x_test, y_test, verbose=0)[1],model2.evaluate(x_test, y_test, verbose=0)[1],model3.evaluate(x_test, y_test, verbose=0)[1],model4.evaluate(x_test, y_test, verbose=0)[1],model5.evaluate(x_test, y_test, verbose=0)[1],model6.evaluate(x_test, y_test, verbose=0)[1],model7.evaluate(x_test, y_test, verbose=0)[1],model8.evaluate(x_test, y_test, verbose=0)[1],model9.evaluate(x_test, y_test, verbose=0)[1]]
```

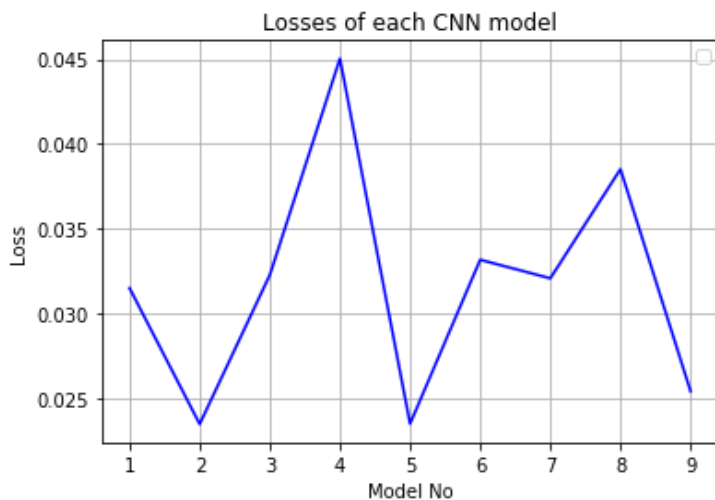
```
In [0]:
```

```
models = list(range(1,10))
```

```
In [0]:
```

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('Model No') ; ax.set_ylabel('Loss')
ax.plot(models, x, 'b')
plt.legend()
plt.grid()
plt.title('Losses of each CNN model')
#plt.show()
fig.canvas.draw()
```

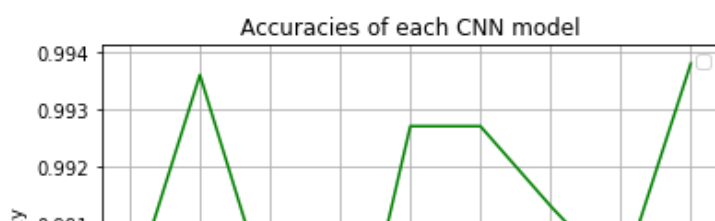
W0825 12:42:55.277995 140359865685888 legend.py:1289] No handles with labels found to put in legend.

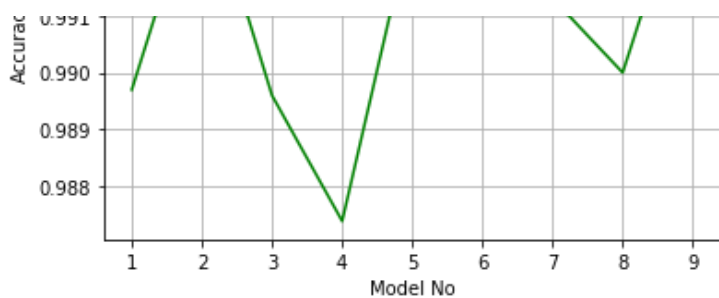


```
In [0]:
```

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('Model No') ; ax.set_ylabel('Accuracy')
ax.plot(models, y, 'g')
plt.legend()
plt.grid()
plt.title('Accuracies of each CNN model')
#plt.show()
fig.canvas.draw()
```

W0825 12:45:15.068250 140359865685888 legend.py:1289] No handles with labels found to put in legend.





10.3 CONVOLUTIONAL LAYERS (MAX POOLING , 2X2 FILTER-RANDOM UNIFORM INITIALISER)

In [5]:

```
#from keras.layers.normalization import BatchNormalization
modell1 = Sequential()

modell1.add(Conv2D(64, kernel_size=(2, 2),
                  activation='relu',
                  input_shape=input_shape,kernel_initializer='random_uniform'))
modell1.add(MaxPooling2D(pool_size=(2, 2)))

modell1.add(Conv2D(32, (2, 2), activation='relu',kernel_initializer='random_uniform'))
modell1.add(MaxPooling2D(pool_size=(2, 2)))

modell1.add(Conv2D(32, (2, 2), activation='relu',kernel_initializer='random_uniform'))
modell1.add(MaxPooling2D(pool_size=(2, 2)))

modell1.add(Flatten())
modell1.add(Dense(128, activation='relu',kernel_initializer='random_uniform'))
modell1.add(Dropout(0.5))
modell1.add(Dense(num_classes, activation='softmax'))

modell1.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = modell1.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = modell1.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 [=====] - 6s 107us/step - loss: 0.8491 - acc: 0.7100
- val_loss: 0.1752 - val_acc: 0.9444
Epoch 2/12
60000/60000 [=====] - 6s 97us/step - loss: 0.2067 - acc: 0.9372
- val_loss: 0.1065 - val_acc: 0.9668
Epoch 3/12
60000/60000 [=====] - 6s 96us/step - loss: 0.1483 - acc: 0.9561
- val_loss: 0.0824 - val_acc: 0.9747
Epoch 4/12
60000/60000 [=====] - 6s 95us/step - loss: 0.1204 - acc: 0.9638
- val_loss: 0.0804 - val_acc: 0.9750
Epoch 5/12
60000/60000 [=====] - 6s 96us/step - loss: 0.1066 - acc: 0.9680
- val_loss: 0.0674 - val_acc: 0.9784
Epoch 6/12
60000/60000 [=====] - 6s 95us/step - loss: 0.0935 - acc: 0.9722
- val_loss: 0.0688 - val_acc: 0.9791
Epoch 7/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0849 - acc: 0.9746
- val_loss: 0.0556 - val_acc: 0.9818
Epoch 8/12
60000/60000 [=====] - 6s 95us/step - loss: 0.0783 - acc: 0.9761
```

```

- val_loss: 0.0541 - val_acc: 0.9844
Epoch 9/12
60000/60000 [=====] - 6s 95us/step - loss: 0.0735 - acc: 0.9780
- val_loss: 0.0511 - val_acc: 0.9834
Epoch 10/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0669 - acc: 0.9798
- val_loss: 0.0495 - val_acc: 0.9852
Epoch 11/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0629 - acc: 0.9812
- val_loss: 0.0522 - val_acc: 0.9837
Epoch 12/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0587 - acc: 0.9816
- val_loss: 0.0449 - val_acc: 0.9862
Test loss: 0.04492342894774338
Test accuracy: 0.9862

```

In [6]:

```
model1.summary()
```

Model: "sequential_3"

| Layer (type) | Output Shape | Param # |
|--------------------------------|--------------------|---------|
| conv2d_4 (Conv2D) | (None, 27, 27, 64) | 320 |
| max_pooling2d_2 (MaxPooling2D) | (None, 13, 13, 64) | 0 |
| conv2d_5 (Conv2D) | (None, 12, 12, 32) | 8224 |
| max_pooling2d_3 (MaxPooling2D) | (None, 6, 6, 32) | 0 |
| conv2d_6 (Conv2D) | (None, 5, 5, 32) | 4128 |
| max_pooling2d_4 (MaxPooling2D) | (None, 2, 2, 32) | 0 |
| flatten_2 (Flatten) | (None, 128) | 0 |
| dense_3 (Dense) | (None, 128) | 16512 |
| dropout_3 (Dropout) | (None, 128) | 0 |
| dense_4 (Dense) | (None, 10) | 1290 |
| Total params: 30,474 | | |
| Trainable params: 30,474 | | |
| Non-trainable params: 0 | | |

In [7]:

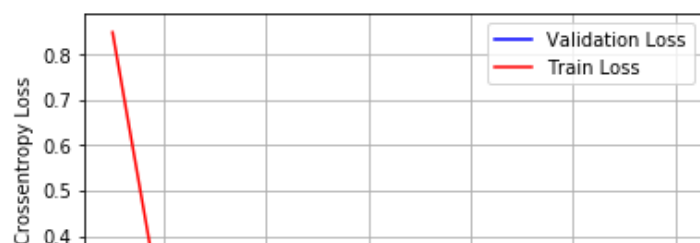
```

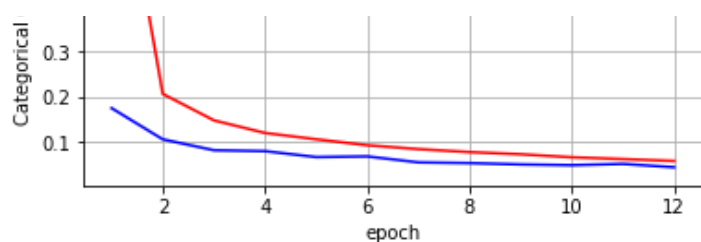
score = model1.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1, epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)

```

Test score: 0.04492342894774338

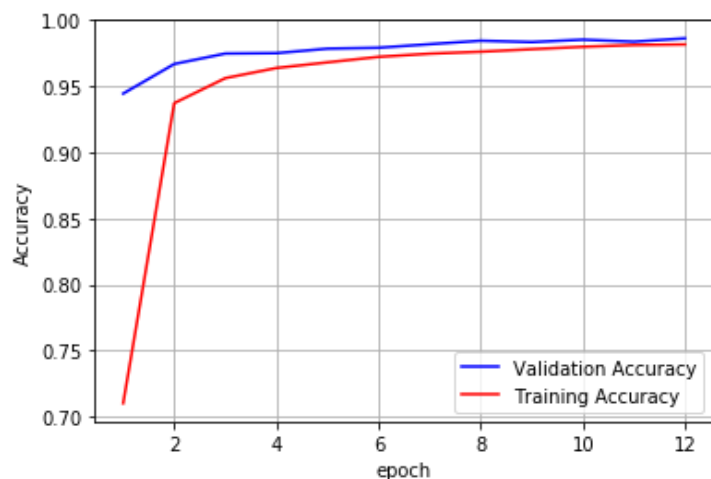
Test accuracy: 0.9862





In [8]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



11.3 CONVOLUTIONAL LAYERS (MAX POOLING , 2X2 FILTER-RANDOM UNIFORM INITIALISER ,TANH ACTIVATION)

In [9]:

```
#from keras.layers.normalization import BatchNormalization
model2 = Sequential()

model2.add(Conv2D(64, kernel_size=(2, 2),
                  activation='tanh',
                  input_shape=input_shape,kernel_initializer='random_uniform'))
model2.add(MaxPooling2D(pool_size=(2, 2)))

model2.add(Conv2D(32, (2, 2), activation='tanh',kernel_initializer='random_uniform'))
model2.add(MaxPooling2D(pool_size=(2, 2)))

model2.add(Conv2D(32, (2, 2), activation='tanh',kernel_initializer='random_uniform'))
model2.add(MaxPooling2D(pool_size=(2, 2)))

model2.add(Flatten())
model2.add(Dense(128, activation='tanh',kernel_initializer='random_uniform'))
model2.add(Dropout(0.5))
model2.add(Dense(num_classes, activation='softmax'))

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

inspect = model2.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    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
60000/60000 [=====] - 7s 114us/step - loss: 0.5136 - acc: 0.8313
- val_loss: 0.1369 - val_acc: 0.9578
Epoch 2/12
60000/60000 [=====] - 6s 103us/step - loss: 0.1358 - acc: 0.9581
- val_loss: 0.0916 - val_acc: 0.9703
Epoch 3/12
60000/60000 [=====] - 6s 100us/step - loss: 0.1038 - acc: 0.9681
- val_loss: 0.0743 - val_acc: 0.9770
Epoch 4/12
60000/60000 [=====] - 6s 101us/step - loss: 0.0898 - acc: 0.9726
- val_loss: 0.0703 - val_acc: 0.9775
Epoch 5/12
60000/60000 [=====] - 6s 102us/step - loss: 0.0776 - acc: 0.9767
- val_loss: 0.0660 - val_acc: 0.9795
Epoch 6/12
60000/60000 [=====] - 6s 100us/step - loss: 0.0711 - acc: 0.9778
- val_loss: 0.0576 - val_acc: 0.9817
Epoch 7/12
60000/60000 [=====] - 6s 98us/step - loss: 0.0649 - acc: 0.9801
- val_loss: 0.0612 - val_acc: 0.9808
Epoch 8/12
60000/60000 [=====] - 6s 98us/step - loss: 0.0607 - acc: 0.9815
- val_loss: 0.0603 - val_acc: 0.9822
Epoch 9/12
60000/60000 [=====] - 6s 97us/step - loss: 0.0574 - acc: 0.9822
- val_loss: 0.0570 - val_acc: 0.9827
Epoch 10/12
60000/60000 [=====] - 6s 97us/step - loss: 0.0525 - acc: 0.9838
- val_loss: 0.0539 - val_acc: 0.9819
Epoch 11/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0494 - acc: 0.9846
- val_loss: 0.0552 - val_acc: 0.9839
Epoch 12/12
60000/60000 [=====] - 6s 96us/step - loss: 0.0477 - acc: 0.9851
- val_loss: 0.0632 - val_acc: 0.9825
Test loss: 0.0631999080858659
Test accuracy: 0.9825

In [10]:

```
model2.summary()
```

Model: "sequential_4"

| Layer (type) | Output Shape | Param # |
|------------------------------|--------------------|---------|
| ===== | | |
| conv2d_7 (Conv2D) | (None, 27, 27, 64) | 320 |
| max_pooling2d_5 (MaxPooling2 | (None, 13, 13, 64) | 0 |
| conv2d_8 (Conv2D) | (None, 12, 12, 32) | 8224 |
| max_pooling2d_6 (MaxPooling2 | (None, 6, 6, 32) | 0 |
| conv2d_9 (Conv2D) | (None, 5, 5, 32) | 4128 |
| max_pooling2d_7 (MaxPooling2 | (None, 2, 2, 32) | 0 |
| flatten_3 (Flatten) | (None, 128) | 0 |
| dense_5 (Dense) | (None, 128) | 16512 |
| dropout_4 (Dropout) | (None, 128) | 0 |
| dense_6 (Dense) | (None, 10) | 1290 |
| ===== | | |

Total params: 30,474
Trainable params: 30,474

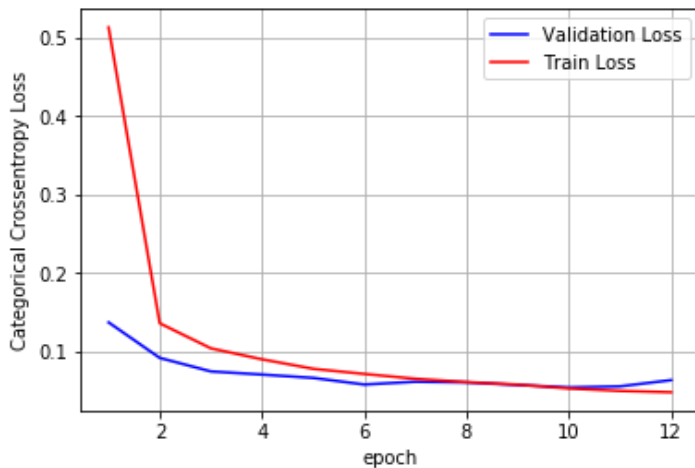
Non-trainable params: 0

In [11]:

```
score = model2.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')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

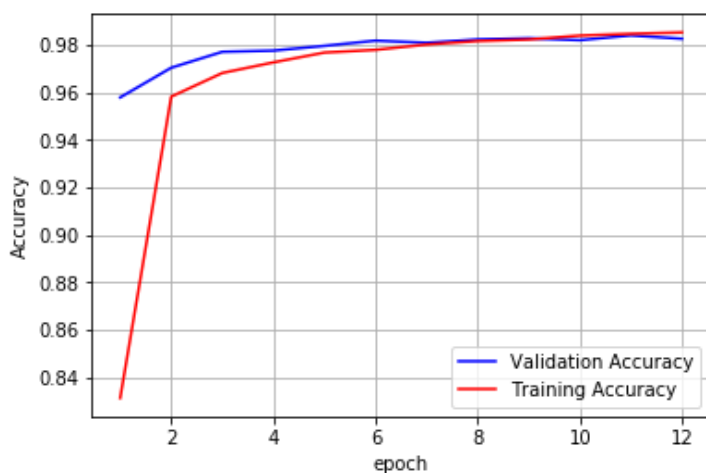
Test score: 0.0631999080858659

Test accuracy: 0.9825



In [12]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



12. 3 CONVOLUTIONAL LAYERS (MAX POOLING ,BATCH NORM,DROPOUT=(0.4) 2X2 FILTER-ORTHOGONAL INITIALISER ,TANH ACTIVATION)

In [18]:

```
from keras.layers.normalization import BatchNormalization
model3 = Sequential()

model3.add(Conv2D(64, kernel_size=(2, 2),
                  activation='tanh',
```

```

        input_shape=input_shape,kernel_initializer='orthogonal'))
model3.add(BatchNormalization())
model3.add(MaxPooling2D(pool_size=(2, 2)))
model3.add(Dropout(0.4))

model3.add(Conv2D(32, (2, 2), activation='tanh',kernel_initializer='orthogonal'))
model3.add(BatchNormalization())
model3.add(MaxPooling2D(pool_size=(2, 2)))
model3.add(Dropout(0.4))

model3.add(Conv2D(32, (2, 2), activation='tanh',kernel_initializer='orthogonal'))
model3.add(BatchNormalization())
model3.add(MaxPooling2D(pool_size=(2, 2)))
model3.add(Dropout(0.4))

model3.add(Flatten())
model3.add(Dense(128, activation='tanh'))
model3.add(Dropout(0.4))
model3.add(Dense(num_classes, activation='softmax'))

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

inspect = model3.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    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 1/12
60000/60000 [=====] - 10s 170us/step - loss: 1.2124 - acc: 0.590
6 - val_loss: 0.4706 - val_acc: 0.8536
Epoch 2/12
60000/60000 [=====] - 9s 150us/step - loss: 0.5745 - acc: 0.8124
- val_loss: 0.5390 - val_acc: 0.8286
Epoch 3/12
60000/60000 [=====] - 9s 150us/step - loss: 0.4348 - acc: 0.8603
- val_loss: 0.2854 - val_acc: 0.9100
Epoch 4/12
60000/60000 [=====] - 9s 150us/step - loss: 0.3666 - acc: 0.8842
- val_loss: 0.1562 - val_acc: 0.9510
Epoch 5/12
60000/60000 [=====] - 9s 152us/step - loss: 0.3253 - acc: 0.8980
- val_loss: 0.1373 - val_acc: 0.9574
Epoch 6/12
60000/60000 [=====] - 9s 152us/step - loss: 0.2971 - acc: 0.9069
- val_loss: 0.1358 - val_acc: 0.9565
Epoch 7/12
60000/60000 [=====] - 9s 151us/step - loss: 0.2755 - acc: 0.9142
- val_loss: 0.1017 - val_acc: 0.9687
Epoch 8/12
60000/60000 [=====] - 9s 156us/step - loss: 0.2551 - acc: 0.9197
- val_loss: 0.0934 - val_acc: 0.9711
Epoch 9/12
60000/60000 [=====] - 9s 152us/step - loss: 0.2426 - acc: 0.9250
- val_loss: 0.0943 - val_acc: 0.9725
Epoch 10/12
60000/60000 [=====] - 9s 154us/step - loss: 0.2354 - acc: 0.9267
- val_loss: 0.0936 - val_acc: 0.9703
Epoch 11/12
60000/60000 [=====] - 9s 152us/step - loss: 0.2192 - acc: 0.9301
- val_loss: 0.0925 - val_acc: 0.9719
Epoch 12/12
60000/60000 [=====] - 9s 152us/step - loss: 0.2121 - acc: 0.9334
- val_loss: 0.0757 - val_acc: 0.9772
Test loss: 0.07570360766109079

```


Test accuracy: 0.9772

In [15]:

```
model3.summary()
```

Model: "sequential_6"

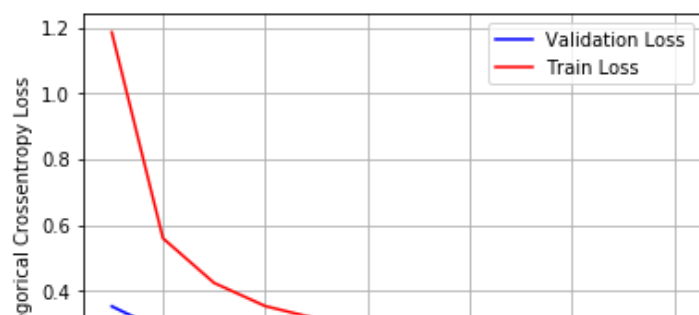
| Layer (type) | Output Shape | Param # |
|---|--------------------|---------|
| conv2d_11 (Conv2D) | (None, 27, 27, 64) | 320 |
| batch_normalization_1 (Batch Normalization) | (None, 27, 27, 64) | 256 |
| max_pooling2d_8 (MaxPooling2D) | (None, 13, 13, 64) | 0 |
| dropout_5 (Dropout) | (None, 13, 13, 64) | 0 |
| conv2d_12 (Conv2D) | (None, 12, 12, 32) | 8224 |
| batch_normalization_2 (Batch Normalization) | (None, 12, 12, 32) | 128 |
| max_pooling2d_9 (MaxPooling2D) | (None, 6, 6, 32) | 0 |
| dropout_6 (Dropout) | (None, 6, 6, 32) | 0 |
| conv2d_13 (Conv2D) | (None, 5, 5, 32) | 4128 |
| batch_normalization_3 (Batch Normalization) | (None, 5, 5, 32) | 128 |
| max_pooling2d_10 (MaxPooling2D) | (None, 2, 2, 32) | 0 |
| dropout_7 (Dropout) | (None, 2, 2, 32) | 0 |
| flatten_4 (Flatten) | (None, 128) | 0 |
| dense_7 (Dense) | (None, 128) | 16512 |
| dropout_8 (Dropout) | (None, 128) | 0 |
| dense_8 (Dense) | (None, 10) | 1290 |
| Total params: 30,986 | | |
| Trainable params: 30,730 | | |
| Non-trainable params: 256 | | |

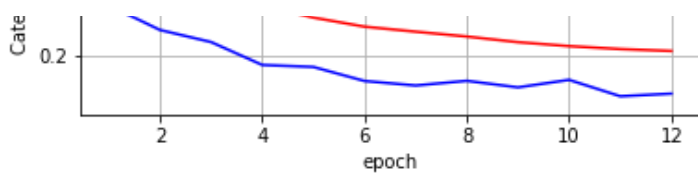
In [16]:

```
score = model3.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')
x = list(range(1, epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)
```

Test score: 0.0859176158150658

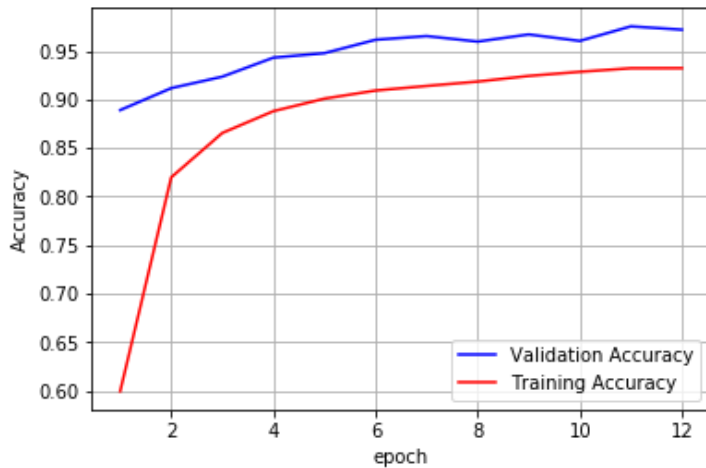
Test accuracy: 0.9718





In [17]:

```
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1, epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



13. 3 CONVOLUTIONAL LAYERS (MAX POOLING, 3X3 FILTER-ORTHOGONAL INITIALISER ,SIGMOID ACTIVATION)

In [20]:

```
from keras.layers.normalization import BatchNormalization
model4 = Sequential()

model4.add(Conv2D(64, kernel_size=(3,3),
                  activation='sigmoid',
                  input_shape=input_shape, kernel_initializer='orthogonal'))
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Conv2D(32, (3, 3), activation='sigmoid', kernel_initializer='orthogonal'))
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Conv2D(32, (3, 3), activation='sigmoid', kernel_initializer='orthogonal'))
model4.add(MaxPooling2D(pool_size=(2, 2)))

model4.add(Flatten())
model4.add(Dense(128, activation='sigmoid'))
model4.add(Dropout(0.4))
model4.add(Dense(num_classes, activation='softmax'))

model4.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adadelta(),
               metrics=['accuracy'])

inspect = model4.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model4.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 [=====] - 8s 126us/step - loss: 2.3453 - acc: 0.1022
- val_loss: 2.3022 - val_acc: 0.1135
Epoch 2/12
60000/60000 [=====] - 6s 103us/step - loss: 2.3064 - acc: 0.1058
- val_loss: 2.3022 - val_acc: 0.1009
Epoch 3/12
60000/60000 [=====] - 6s 101us/step - loss: 2.3038 - acc: 0.1077
- val_loss: 2.3013 - val_acc: 0.1135
Epoch 4/12
60000/60000 [=====] - 6s 99us/step - loss: 2.3024 - acc: 0.1097
- val_loss: 2.3012 - val_acc: 0.1135
Epoch 5/12
60000/60000 [=====] - 6s 99us/step - loss: 2.3021 - acc: 0.1108
- val_loss: 2.3013 - val_acc: 0.1135
Epoch 6/12
60000/60000 [=====] - 6s 102us/step - loss: 2.3021 - acc: 0.1113
- val_loss: 2.3011 - val_acc: 0.1135
Epoch 7/12
60000/60000 [=====] - 6s 104us/step - loss: 2.3016 - acc: 0.1118
- val_loss: 2.3015 - val_acc: 0.1135
Epoch 8/12
60000/60000 [=====] - 6s 101us/step - loss: 2.3018 - acc: 0.1113
- val_loss: 2.3011 - val_acc: 0.1135
Epoch 9/12
60000/60000 [=====] - 6s 104us/step - loss: 2.3016 - acc: 0.1121
- val_loss: 2.3014 - val_acc: 0.1135
Epoch 10/12
60000/60000 [=====] - 6s 100us/step - loss: 2.3015 - acc: 0.1121
- val_loss: 2.3014 - val_acc: 0.1135
Epoch 11/12
60000/60000 [=====] - 6s 100us/step - loss: 2.3017 - acc: 0.1125
- val_loss: 2.3010 - val_acc: 0.1135
Epoch 12/12
60000/60000 [=====] - 6s 100us/step - loss: 2.3017 - acc: 0.1125
- val_loss: 2.3010 - val_acc: 0.1135
Test loss: 2.3009641578674316
Test accuracy: 0.1135

In [21]:

```
model4.summary()
```

Model: "sequential_9"

| Layer (type) | Output Shape | Param # |
|-------------------------------|--------------------|---------|
| ===== | | |
| conv2d_20 (Conv2D) | (None, 26, 26, 64) | 640 |
| max_pooling2d_17 (MaxPooling) | (None, 13, 13, 64) | 0 |
| conv2d_21 (Conv2D) | (None, 11, 11, 32) | 18464 |
| max_pooling2d_18 (MaxPooling) | (None, 5, 5, 32) | 0 |
| conv2d_22 (Conv2D) | (None, 3, 3, 32) | 9248 |
| max_pooling2d_19 (MaxPooling) | (None, 1, 1, 32) | 0 |
| flatten_7 (Flatten) | (None, 32) | 0 |
| dense_13 (Dense) | (None, 128) | 4224 |
| dropout_14 (Dropout) | (None, 128) | 0 |
| dense_14 (Dense) | (None, 10) | 1290 |
| ===== | | |

Total params: 33,866
Trainable params: 33,866

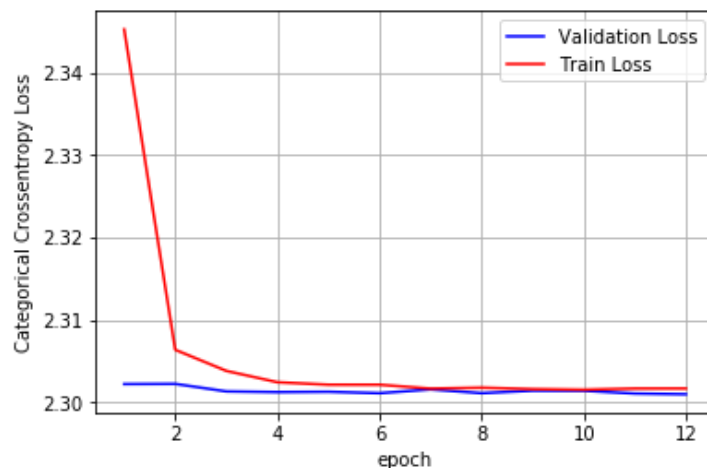
Non-trainable params: 0

In [22]:

```
score = model4.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')
x = list(range(1,epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

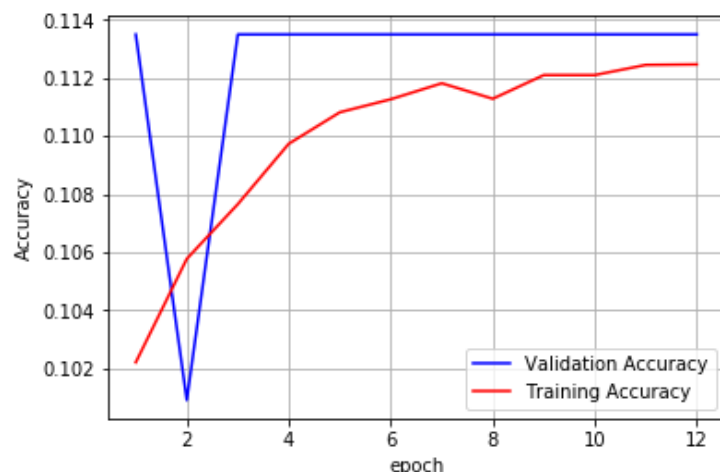
Test score: 2.3009641578674316

Test accuracy: 0.1135



In [23]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



14. 5 CONVOLUTIONAL LAYERS (MAX POOLING AND BATCH NORMALIZATION 3X3 FILTER,ELU ACTIVATION,RMSProp OPTIMISER)

In [27]:

```
from keras.layers.normalization import BatchNormalization
model5 = Sequential()
#1ST
model5.add(Conv2D(256, kernel_size=(3, 3),
                  activation='elu',
```

```

        input_shape=input_shape,padding="same",kernel_initializer='zeros'))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.6))

#2ND
model5.add(Conv2D(128, (3, 3), activation='elu',padding="same",kernel_initializer='zeros'
))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#3RD
model5.add(Conv2D(64, (3, 3), activation='elu',padding="same",kernel_initializer='zeros'
))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#4TH
model5.add(Conv2D(32, (3, 3), activation='elu',padding="same",kernel_initializer='zeros'
))
model5.add(BatchNormalization())
model5.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#5TH
model5.add(Conv2D(16, (3, 3), activation='elu',padding="same",kernel_initializer='zeros'
))
model5.add(BatchNormalization())
#model5.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

model5.add(Flatten())
model5.add(Dense(32, activation='elu'))

model5.add(Dense(num_classes, activation='softmax'))

model5.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.RMSprop(),
               metrics=['accuracy'])

inspect = model5.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model5.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 [=====] - 30s 505us/step - loss: 2.3017 - acc: 0.112
0 - val_loss: 2.3062 - val_acc: 0.1135

Epoch 2/12

60000/60000 [=====] - 28s 468us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3040 - val_acc: 0.1135

Epoch 3/12

60000/60000 [=====] - 28s 471us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3042 - val_acc: 0.1135

Epoch 4/12

60000/60000 [=====] - 28s 470us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3038 - val_acc: 0.1032

Epoch 5/12

60000/60000 [=====] - 28s 473us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3033 - val_acc: 0.1135

Epoch 6/12

60000/60000 [=====] - 28s 470us/step - loss: 2.3014 - acc: 0.112
3 - val_loss: 2.3017 - val_acc: 0.1135

```

Epoch 7/12
60000/60000 [=====] - 28s 468us/step - loss: 2.3014 - acc: 0.112
0 - val_loss: 2.3025 - val_acc: 0.1135
Epoch 8/12
60000/60000 [=====] - 28s 474us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3012 - val_acc: 0.1135
Epoch 9/12
60000/60000 [=====] - 28s 468us/step - loss: 2.3015 - acc: 0.112
4 - val_loss: 2.3012 - val_acc: 0.1135
Epoch 10/12
60000/60000 [=====] - 28s 474us/step - loss: 2.3014 - acc: 0.112
4 - val_loss: 2.3015 - val_acc: 0.1135
Epoch 11/12
60000/60000 [=====] - 28s 474us/step - loss: 2.3014 - acc: 0.112
4 - val_loss: 2.3011 - val_acc: 0.1135
Epoch 12/12
60000/60000 [=====] - 28s 473us/step - loss: 2.3014 - acc: 0.112
4 - val_loss: 2.3017 - val_acc: 0.1135
Test loss: 2.3016848217010497
Test accuracy: 0.1135

```

In [28]:

```
model5.summary()
```

Model: "sequential_13"

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_33 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_15 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_27 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| conv2d_34 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_16 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_28 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| conv2d_35 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_17 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_29 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| conv2d_36 (Conv2D) | (None, 3, 3, 32) | 18464 |
| batch_normalization_18 (Batch Normalization) | (None, 3, 3, 32) | 128 |
| max_pooling2d_30 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| conv2d_37 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_19 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| flatten_9 (Flatten) | (None, 16) | 0 |
| dense_17 (Dense) | (None, 32) | 544 |
| dense_18 (Dense) | (None, 10) | 330 |
| ===== | | |
| Total params: 397,338 | | |
| Trainable params: 396,346 | | |
| Non-trainable params: 992 | | |

In [29]:

```

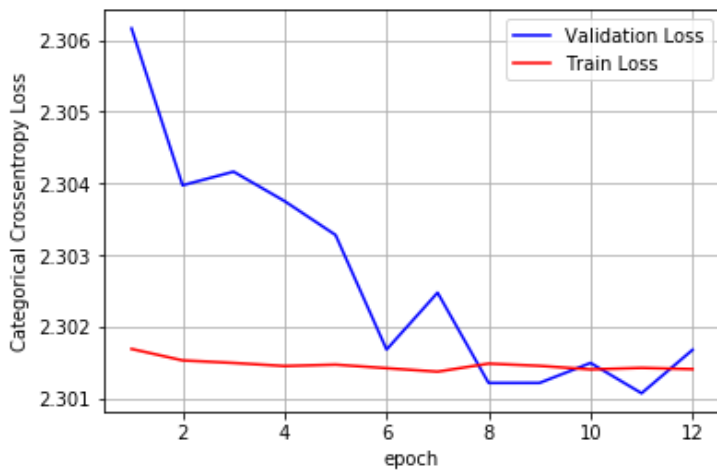
score = model5.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])

```

```
print('Test accuracy:', score[1])
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1,epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)
```

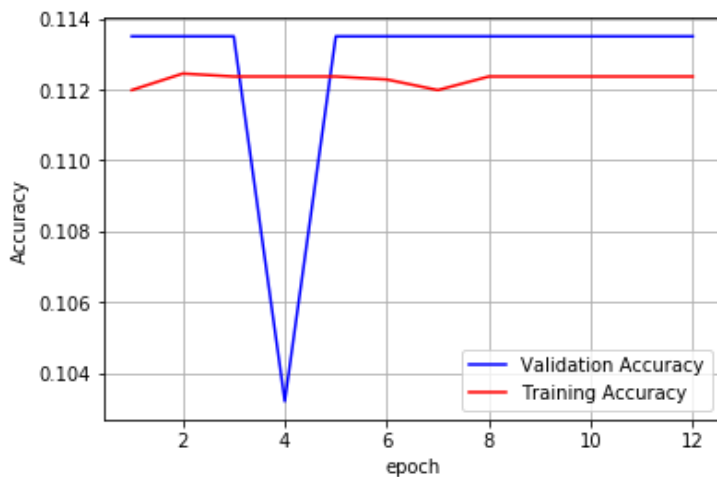
Test score: 2.3016848217010497

Test accuracy: 0.1135



In [30]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



15. 5 CONVOLUTIONAL LAYERS (MAX POOLING,DROPOUT AND BATCH NORMALIZATION 3X3 FILTER,RMSProp OPTIMISER, TANH Activation)

In [32]:

```
from keras.layers.normalization import BatchNormalization
model6 = Sequential()
#1ST
model6.add(Conv2D(256, kernel_size=(3, 3),
                  activation='tanh',
                  input_shape=input_shape,padding="same",kernel_initializer='ones'))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.6))

#2ND
```

```

model6.add(Conv2D(128, (3, 3), activation='tanh',padding="same",kernel_initializer='ones'
))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#3RD
model6.add(Conv2D(64, (3, 3), activation='tanh',padding="same",kernel_initializer='ones'
))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#4TH
model6.add(Conv2D(32, (3, 3), activation='tanh',padding="same",kernel_initializer='ones'
))
model6.add(BatchNormalization())
model6.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#5TH
model6.add(Conv2D(16, (3, 3), activation='tanh',padding="same",kernel_initializer='ones'
))
model6.add(BatchNormalization())
#model.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

model6.add(Flatten())
model6.add(Dense(32, activation='tanh'))

model6.add(Dense(num_classes, activation='softmax'))

model6.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.RMSprop(),
               metrics=['accuracy'])

inspect = model6.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model6.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 [=====] - 30s 505us/step - loss: 2.1452 - acc: 0.181
9 - val_loss: 2.1244 - val_acc: 0.1849
Epoch 2/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1422 - acc: 0.183
4 - val_loss: 2.1243 - val_acc: 0.1903
Epoch 3/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1416 - acc: 0.182
9 - val_loss: 2.1280 - val_acc: 0.1857
Epoch 4/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1412 - acc: 0.181
8 - val_loss: 2.1236 - val_acc: 0.1850
Epoch 5/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1404 - acc: 0.184
4 - val_loss: 2.1242 - val_acc: 0.1903
Epoch 6/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1405 - acc: 0.181
3 - val_loss: 2.1249 - val_acc: 0.1856
Epoch 7/12
60000/60000 [=====] - 28s 466us/step - loss: 2.1403 - acc: 0.181
9 - val_loss: 2.1243 - val_acc: 0.1856
Epoch 8/12
60000/60000 [=====] - 28s 466us/step - loss: 2.1406 - acc: 0.181
9 - val_loss: 2.1232 - val_acc: 0.1849
Epoch 9/12

```



```

60000/60000 [=====] - 28s 468us/step - loss: 2.1405 - acc: 0.181
7 - val_loss: 2.1232 - val_acc: 0.1857
Epoch 10/12
60000/60000 [=====] - 28s 463us/step - loss: 2.1405 - acc: 0.182
9 - val_loss: 2.1241 - val_acc: 0.1856
Epoch 11/12
60000/60000 [=====] - 28s 460us/step - loss: 2.1406 - acc: 0.183
1 - val_loss: 2.1253 - val_acc: 0.1857
Epoch 12/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1411 - acc: 0.180
9 - val_loss: 2.1242 - val_acc: 0.1857
Test loss: 2.1241648181915282
Test accuracy: 0.1857

```

In [33]:

```
model6.summary()
```

Model: "sequential_15"

| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| conv2d_42 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_23 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_34 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| conv2d_43 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_24 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_35 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| conv2d_44 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_25 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_36 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| conv2d_45 (Conv2D) | (None, 3, 3, 32) | 18464 |
| batch_normalization_26 (Batch Normalization) | (None, 3, 3, 32) | 128 |
| max_pooling2d_37 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| conv2d_46 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_27 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| flatten_10 (Flatten) | (None, 16) | 0 |
| dense_19 (Dense) | (None, 32) | 544 |
| dense_20 (Dense) | (None, 10) | 330 |
| Total params: 397,338 | | |
| Trainable params: 396,346 | | |
| Non-trainable params: 992 | | |

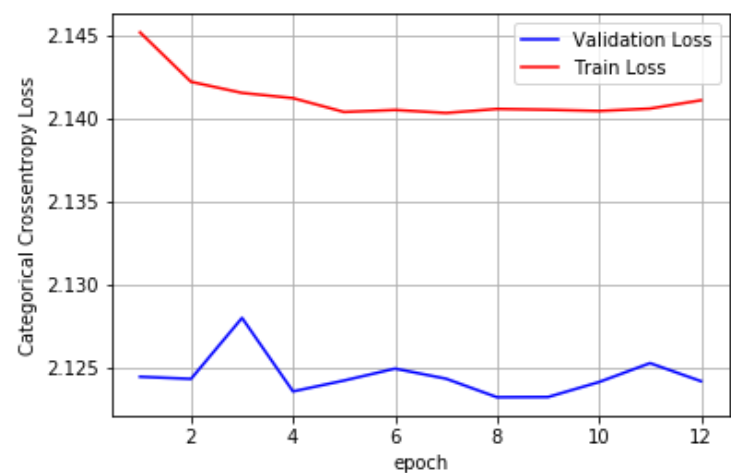
In [34]:

```

score = model6.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')
x = list(range(1, epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)

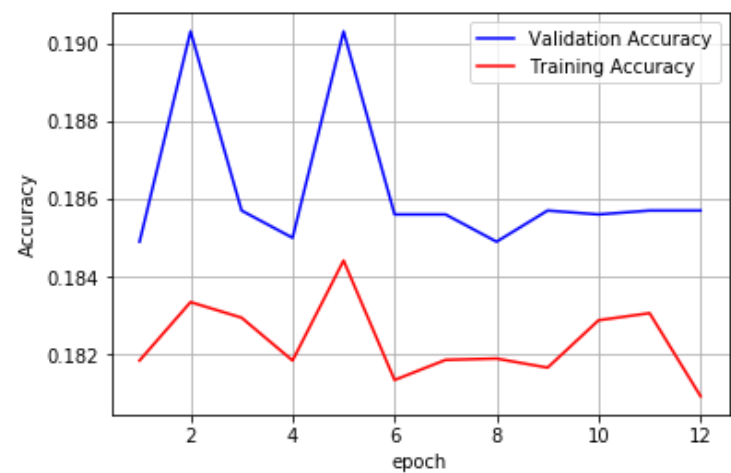
```

Test score: 2.1241648181915282
Test accuracy: 0.1857



In [35]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



16. 5 CONVOLUTIONAL LAYERS (MAX POOLING, 3X3 FILTER,RMSProp OPTIMISER,TANH Activation,RANDOM UNIFORM INITIALISER)

In [37]:

```
from keras.layers.normalization import BatchNormalization
model7 = Sequential()
#1ST
model7.add(Conv2D(256, kernel_size=(3, 3),
                  activation='tanh',
                  input_shape=input_shape,padding="same",kernel_initializer='random_unifor
rm'))
#model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.6))

#2ND
model7.add(Conv2D(128, (3, 3), activation='tanh',padding="same",kernel_initializer='rand
om_uniform'))

#model6.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))
```

```

#3RD
model7.add(Conv2D(64, (3, 3), activation='tanh',padding="same",kernel_initializer='random_uniform'))
#model7.add(BatchNormalization())
model7.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#4TH
model7.add(Conv2D(32, (3, 3), activation='tanh',padding="same",kernel_initializer='random_uniform'))
#model7.add(BatchNormalization())
#model7.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

#5TH
model7.add(Conv2D(16, (3, 3), activation='tanh',padding="same",kernel_initializer='random_uniform'))
#model6.add(BatchNormalization())
#model7.add(MaxPooling2D(pool_size=(2, 2)))
#model5.add(Dropout(0.5))

model7.add(Flatten())
model7.add(Dense(32, activation='tanh'))

model7.add(Dense(num_classes, activation='softmax'))

model7.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.RMSprop(),
               metrics=['accuracy'])

inspect = model7.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model7.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 [=====] - 29s 483us/step - loss: 2.1406 - acc: 0.182
4 - val_loss: 2.1229 - val_acc: 0.1850
Epoch 2/12
60000/60000 [=====] - 28s 463us/step - loss: 2.1403 - acc: 0.181
2 - val_loss: 2.1250 - val_acc: 0.1903
Epoch 3/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1406 - acc: 0.182
6 - val_loss: 2.1245 - val_acc: 0.1856
Epoch 4/12
60000/60000 [=====] - 28s 467us/step - loss: 2.1403 - acc: 0.181
9 - val_loss: 2.1243 - val_acc: 0.1856
Epoch 5/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1398 - acc: 0.183
2 - val_loss: 2.1243 - val_acc: 0.1849
Epoch 6/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1408 - acc: 0.183
4 - val_loss: 2.1240 - val_acc: 0.1856
Epoch 7/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1401 - acc: 0.182
9 - val_loss: 2.1243 - val_acc: 0.1858
Epoch 8/12
60000/60000 [=====] - 28s 465us/step - loss: 2.1397 - acc: 0.182
8 - val_loss: 2.1246 - val_acc: 0.1856
Epoch 9/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1404 - acc: 0.184
2 - val_loss: 2.1232 - val_acc: 0.1857
Epoch 10/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1406 - acc: 0.184
2 - val_loss: 2.1243 - val_acc: 0.1903

```

```
Epoch 11/12
60000/60000 [=====] - 28s 462us/step - loss: 2.1409 - acc: 0.182
9 - val_loss: 2.1257 - val_acc: 0.1850
Epoch 12/12
60000/60000 [=====] - 28s 464us/step - loss: 2.1400 - acc: 0.181
6 - val_loss: 2.1231 - val_acc: 0.1849
Test loss: 2.123111280250549
Test accuracy: 0.1849
```

In [38]:

```
model7.summary()
```

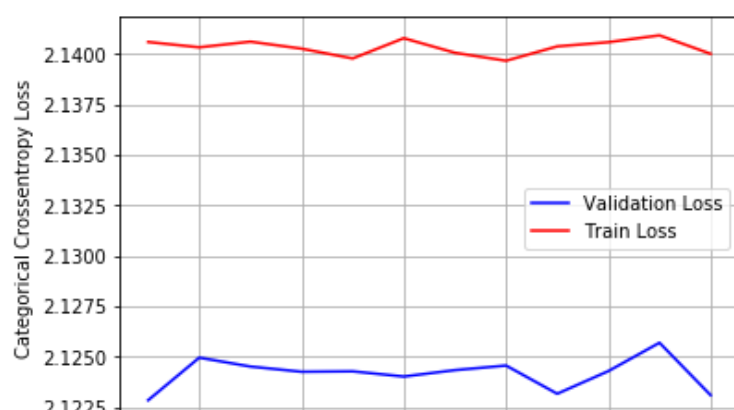
Model: "sequential_17"

| Layer (type) | Output Shape | Param # |
|-------------------------------|---------------------|---------|
| conv2d_52 (Conv2D) | (None, 28, 28, 256) | 2560 |
| max_pooling2d_43 (MaxPooling) | (None, 14, 14, 256) | 0 |
| conv2d_53 (Conv2D) | (None, 14, 14, 128) | 295040 |
| max_pooling2d_44 (MaxPooling) | (None, 7, 7, 128) | 0 |
| conv2d_54 (Conv2D) | (None, 7, 7, 64) | 73792 |
| max_pooling2d_45 (MaxPooling) | (None, 3, 3, 64) | 0 |
| conv2d_55 (Conv2D) | (None, 3, 3, 32) | 18464 |
| conv2d_56 (Conv2D) | (None, 3, 3, 16) | 4624 |
| flatten_11 (Flatten) | (None, 144) | 0 |
| dense_21 (Dense) | (None, 32) | 4640 |
| dense_22 (Dense) | (None, 10) | 330 |
| Total params: 399,450 | | |
| Trainable params: 399,450 | | |
| Non-trainable params: 0 | | |

In [39]:

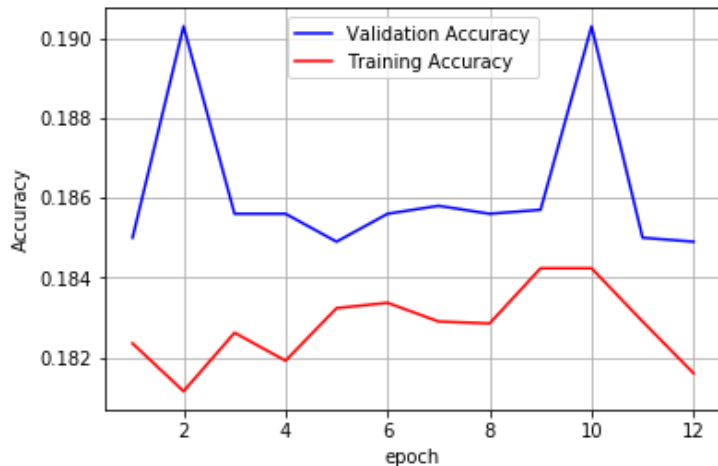
```
score = model7.evaluate(x_test, y_test, verbose=0)
print('Test score:', score[0])
print('Test accuracy:', score[1])
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Categorical Crossentropy Loss')
x = list(range(1, epochs+1))
vl = inspect.history['val_loss']
tl = inspect.history['loss']
plt_dynamic(x, vl, tl, ax)
```

Test score: 2.3014822479248047
Test accuracy: 0.0977



In [40]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



17. 5 CONVOLUTIONAL LAYERS (MAX POOLING,BATCHNORM AND DROPOUT(0.5 and 0.6) 3X3 FILTER,SGD OPTIMISER,RELU Activation,RANDOM UNIFORM INITIALISER)

In [43]:

```
from keras.layers.normalization import BatchNormalization
model8 = Sequential()
#1ST
model8.add(Conv2D(256, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape,padding="same",kernel_initializer='random_unifor
rm'))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.6))

#2ND
model8.add(Conv2D(128, (3, 3), activation='relu',padding="same",kernel_initializer='rand
om_uniform'))

model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#3RD
model8.add(Conv2D(64, (3, 3), activation='relu',padding="same",kernel_initializer='rando
m_uniform'))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#4TH
model8.add(Conv2D(32, (3, 3), activation='relu',padding="same",kernel_initializer='rando
m_uniform'))
model8.add(BatchNormalization())
model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

#5TH
```

```

model8.add(Conv2D(16, (3, 3), activation='relu',padding="same",kernel_initializer='random_uniform'))
model8.add(BatchNormalization())
#model8.add(MaxPooling2D(pool_size=(2, 2)))
model8.add(Dropout(0.5))

model8.add(Flatten())
model8.add(Dense(32, activation='relu'))

model8.add(Dense(num_classes, activation='softmax'))

model8.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.SGD(),
               metrics=['accuracy'])

inspect = model8.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model8.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])

```

WARNING:tensorflow:Large dropout rate: 0.6 (>0.5). In TensorFlow 2.x, dropout() uses dropout rate instead of keep_prob. Please ensure that this is intended.

Train on 60000 samples, validate on 10000 samples

```

Epoch 1/12
60000/60000 [=====] - 33s 544us/step - loss: 2.2076 - acc: 0.1946 - val_loss: 2.9449 - val_acc: 0.1193
Epoch 2/12
60000/60000 [=====] - 30s 497us/step - loss: 1.7290 - acc: 0.3785 - val_loss: 1.9839 - val_acc: 0.3266
Epoch 3/12
60000/60000 [=====] - 30s 497us/step - loss: 1.4061 - acc: 0.4878 - val_loss: 1.3686 - val_acc: 0.5544
Epoch 4/12
60000/60000 [=====] - 30s 505us/step - loss: 1.1591 - acc: 0.5782 - val_loss: 0.8924 - val_acc: 0.7431
Epoch 5/12
60000/60000 [=====] - 30s 503us/step - loss: 0.9946 - acc: 0.6436 - val_loss: 0.6976 - val_acc: 0.7971
Epoch 6/12
60000/60000 [=====] - 30s 505us/step - loss: 0.8827 - acc: 0.6832 - val_loss: 0.5779 - val_acc: 0.8454
Epoch 7/12
60000/60000 [=====] - 30s 503us/step - loss: 0.7983 - acc: 0.7183 - val_loss: 0.4070 - val_acc: 0.8990
Epoch 8/12
60000/60000 [=====] - 30s 504us/step - loss: 0.7332 - acc: 0.7447 - val_loss: 0.3589 - val_acc: 0.9148
Epoch 9/12
60000/60000 [=====] - 30s 504us/step - loss: 0.6853 - acc: 0.7600 - val_loss: 0.3746 - val_acc: 0.8940
Epoch 10/12
60000/60000 [=====] - 30s 505us/step - loss: 0.6431 - acc: 0.7780 - val_loss: 0.2435 - val_acc: 0.9396
Epoch 11/12
60000/60000 [=====] - 30s 503us/step - loss: 0.6061 - acc: 0.7921 - val_loss: 0.2162 - val_acc: 0.9463
Epoch 12/12
60000/60000 [=====] - 30s 504us/step - loss: 0.5743 - acc: 0.8042 - val_loss: 0.1933 - val_acc: 0.9499
Test loss: 0.1933117022037506
Test accuracy: 0.9499

```

In [45]:

```
model8.summary()
```

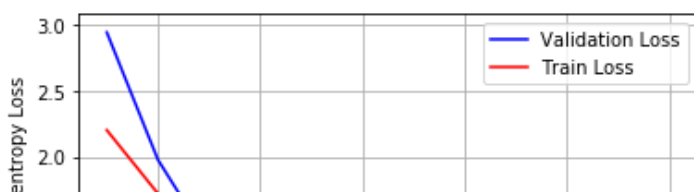
Model: "sequential_20"

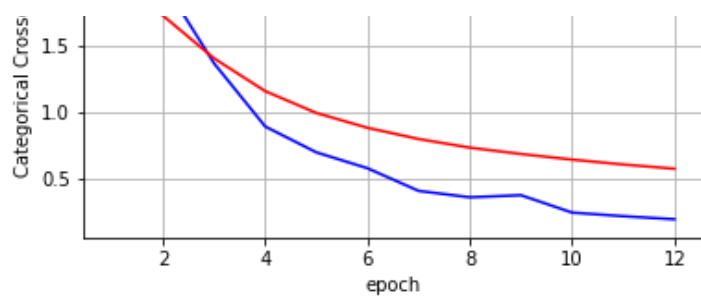
| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| conv2d_67 (Conv2D) | (None, 28, 28, 256) | 2560 |
| batch_normalization_38 (Batch Normalization) | (None, 28, 28, 256) | 1024 |
| max_pooling2d_55 (MaxPooling2D) | (None, 14, 14, 256) | 0 |
| dropout_32 (Dropout) | (None, 14, 14, 256) | 0 |
| conv2d_68 (Conv2D) | (None, 14, 14, 128) | 295040 |
| batch_normalization_39 (Batch Normalization) | (None, 14, 14, 128) | 512 |
| max_pooling2d_56 (MaxPooling2D) | (None, 7, 7, 128) | 0 |
| dropout_33 (Dropout) | (None, 7, 7, 128) | 0 |
| conv2d_69 (Conv2D) | (None, 7, 7, 64) | 73792 |
| batch_normalization_40 (Batch Normalization) | (None, 7, 7, 64) | 256 |
| max_pooling2d_57 (MaxPooling2D) | (None, 3, 3, 64) | 0 |
| dropout_34 (Dropout) | (None, 3, 3, 64) | 0 |
| conv2d_70 (Conv2D) | (None, 3, 3, 32) | 18464 |
| batch_normalization_41 (Batch Normalization) | (None, 3, 3, 32) | 128 |
| max_pooling2d_58 (MaxPooling2D) | (None, 1, 1, 32) | 0 |
| dropout_35 (Dropout) | (None, 1, 1, 32) | 0 |
| conv2d_71 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_42 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| dropout_36 (Dropout) | (None, 1, 1, 16) | 0 |
| flatten_13 (Flatten) | (None, 16) | 0 |
| dense_25 (Dense) | (None, 32) | 544 |
| dense_26 (Dense) | (None, 10) | 330 |
| Total params: 397,338 | | |
| Trainable params: 396,346 | | |
| Non-trainable params: 992 | | |

In [46]:

```
score = model8.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')
x = list(range(1, epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)
```

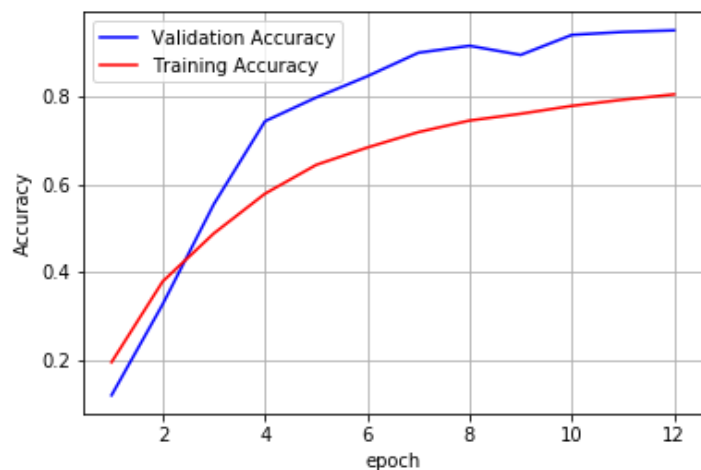
Test score: 0.1933117022037506
Test accuracy: 0.9499





In [47]:

```
fig, ax = plt.subplots(1, 1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1, epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



18. 5 CONVOLUTIONAL LAYERS (MAX POOLING,DROPOUT, 3X3 FILTER,Adamax OPTIMISER,TANH Activation,Orthogonal INITIALISER)

In [48]:

```
from keras.layers.normalization import BatchNormalization
model9 = Sequential()
#1ST
model9.add(Conv2D(256, kernel_size=(3, 3),
                  activation='relu',
                  input_shape=input_shape,padding="same",kernel_initializer='random_uniform'))
#model18.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.3))

#2ND
model9.add(Conv2D(128, (3, 3), activation='relu',padding="same",kernel_initializer='random_uniform'))
#model18.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.3))

#3RD
model9.add(Conv2D(64, (3, 3), activation='relu',padding="same",kernel_initializer='random_uniform'))
#model18.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.3))

#4TH
```



```

model9.add(Conv2D(32, (3, 3), activation='relu',padding="same",kernel_initializer='random_uniform'))
#model8.add(BatchNormalization())
model9.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.3))

#5TH
model9.add(Conv2D(16, (3, 3), activation='relu',padding="same",kernel_initializer='random_uniform'))
model9.add(BatchNormalization())
#model8.add(MaxPooling2D(pool_size=(2, 2)))
model9.add(Dropout(0.3))

model9.add(Flatten())
model9.add(Dense(32, activation='relu'))

model9.add(Dense(num_classes, activation='softmax'))

model9.compile(loss=keras.losses.categorical_crossentropy,
               optimizer=keras.optimizers.Adamax(),
               metrics=['accuracy'])

inspect = model9.fit(x_train, y_train,
                    batch_size=batch_size,
                    epochs=epochs,
                    verbose=1,
                    validation_data=(x_test, y_test))
score = model9.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 [=====] - 25s 414us/step - loss: 1.0174 - acc: 0.660
7 - val_loss: 0.1634 - val_acc: 0.9638
Epoch 2/12
60000/60000 [=====] - 22s 375us/step - loss: 0.2846 - acc: 0.922
4 - val_loss: 0.0631 - val_acc: 0.9820
Epoch 3/12
60000/60000 [=====] - 23s 375us/step - loss: 0.1986 - acc: 0.945
6 - val_loss: 0.0465 - val_acc: 0.9864
Epoch 4/12
60000/60000 [=====] - 23s 378us/step - loss: 0.1543 - acc: 0.958
2 - val_loss: 0.0371 - val_acc: 0.9893
Epoch 5/12
60000/60000 [=====] - 22s 373us/step - loss: 0.1346 - acc: 0.964
1 - val_loss: 0.0367 - val_acc: 0.9903
Epoch 6/12
60000/60000 [=====] - 22s 375us/step - loss: 0.1244 - acc: 0.967
5 - val_loss: 0.0391 - val_acc: 0.9895
Epoch 7/12
60000/60000 [=====] - 22s 372us/step - loss: 0.1139 - acc: 0.969
8 - val_loss: 0.0314 - val_acc: 0.9907
Epoch 8/12
60000/60000 [=====] - 22s 373us/step - loss: 0.1030 - acc: 0.971
4 - val_loss: 0.0270 - val_acc: 0.9925
Epoch 9/12
60000/60000 [=====] - 22s 370us/step - loss: 0.0973 - acc: 0.974
3 - val_loss: 0.0276 - val_acc: 0.9917
Epoch 10/12
60000/60000 [=====] - 22s 368us/step - loss: 0.0952 - acc: 0.975
8 - val_loss: 0.0274 - val_acc: 0.9920
Epoch 11/12
60000/60000 [=====] - 22s 369us/step - loss: 0.0894 - acc: 0.976
6 - val_loss: 0.0249 - val_acc: 0.9927
Epoch 12/12
60000/60000 [=====] - 22s 370us/step - loss: 0.0891 - acc: 0.976
4 - val_loss: 0.0231 - val_acc: 0.9931
Test loss: 0.023112671578422304
Test accuracy: 0.9931

```

In [49]:

```
model9.summary()
```

Model: "sequential_21"

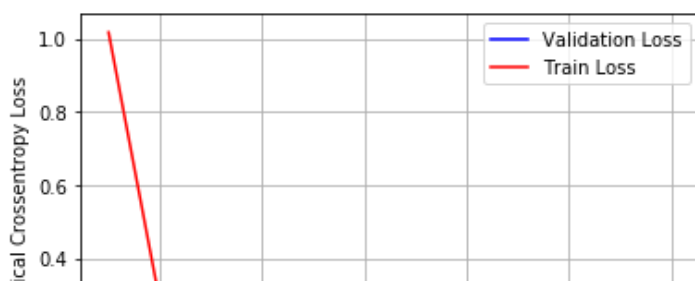
| Layer (type) | Output Shape | Param # |
|--|---------------------|---------|
| ===== | | |
| conv2d_72 (Conv2D) | (None, 28, 28, 256) | 2560 |
| max_pooling2d_59 (MaxPooling) | (None, 14, 14, 256) | 0 |
| dropout_37 (Dropout) | (None, 14, 14, 256) | 0 |
| conv2d_73 (Conv2D) | (None, 14, 14, 128) | 295040 |
| max_pooling2d_60 (MaxPooling) | (None, 7, 7, 128) | 0 |
| dropout_38 (Dropout) | (None, 7, 7, 128) | 0 |
| conv2d_74 (Conv2D) | (None, 7, 7, 64) | 73792 |
| max_pooling2d_61 (MaxPooling) | (None, 3, 3, 64) | 0 |
| dropout_39 (Dropout) | (None, 3, 3, 64) | 0 |
| conv2d_75 (Conv2D) | (None, 3, 3, 32) | 18464 |
| max_pooling2d_62 (MaxPooling) | (None, 1, 1, 32) | 0 |
| dropout_40 (Dropout) | (None, 1, 1, 32) | 0 |
| conv2d_76 (Conv2D) | (None, 1, 1, 16) | 4624 |
| batch_normalization_43 (Batch Normalization) | (None, 1, 1, 16) | 64 |
| dropout_41 (Dropout) | (None, 1, 1, 16) | 0 |
| flatten_14 (Flatten) | (None, 16) | 0 |
| dense_27 (Dense) | (None, 32) | 544 |
| dense_28 (Dense) | (None, 10) | 330 |
| ===== | | |
| Total params: 395,418 | | |
| Trainable params: 395,386 | | |
| Non-trainable params: 32 | | |

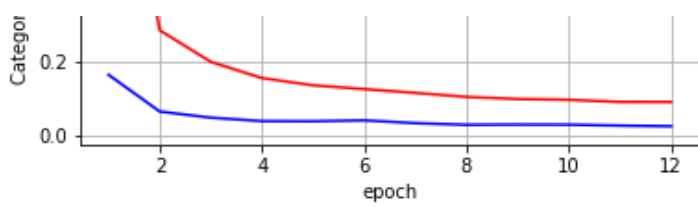
In [50]:

```
score = model9.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')
x = list(range(1, epochs+1))
v1 = inspect.history['val_loss']
t1 = inspect.history['loss']
plt_dynamic(x, v1, t1, ax)
```

Test score: 0.023112671578422304

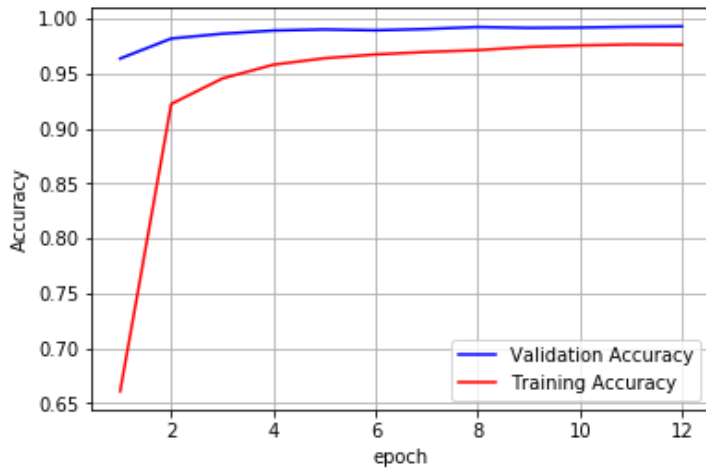
Test accuracy: 0.9931





In [51]:

```
fig,ax = plt.subplots(1,1)
ax.set_xlabel('epoch') ; ax.set_ylabel('Accuracy')
x = list(range(1,epochs+1))
vacc = inspect.history['val_acc']
tacc = inspect.history['acc']
plt_dynamic1(x, vacc, tacc, ax)
```



Conclusions

In [0]:

```
from prettytable import PrettyTable
f='Not present'
t='Present'
t25='Present-(Rate=0.25)'
t5='Present-(Rate=0.5)'
t2='Present-(Rate=0.2)'
test_scr=[0.031,0.023,0.032,0.045,0.023,0.033,0.032,0.038,0.025]
test_acc=[0.9897,0.9936,0.9896,0.9874,0.9927,0.9927,0.9913,0.99,0.9938]
d=[t25,f,t5,f,t5,f,f,t5,t2]
b=[f,t,t,t,t,t,t,t,t]
l=[2,2,2,3,3,3,5,5,7]
sno =[1,2,3,4,5,6,7,8,9]
```

In [0]:

```
table = PrettyTable()
table.add_column('S-NO',sno)
table.add_column("No of Convolution Layers",l)
table.add_column("Dropout Present",d)
table.add_column("Batch Norm Present",b)
table.add_column("Test Loss",test_scr)
table.add_column("Test Acuracy",test_acc)
```

In [0]:

```
print('In the pretty table Dropout is considered only of the convolutional layers.Some models are having a dense layer with dropout value 0.5 and some doesnt have any dropout rate to their dense layer')
print(table)
```

In the pretty table Dropout is considered only of the convolutional layers. Some models are having a dense layer with dropout value 0.5 and some doesn't have any dropout rate to their dense layer

| CNN dense layer | | | | | | |
|----------------------------|--------------------------|---|----------------------|--|--------------------|-----------|
| -----+-----+-----+----- | | | | | | |
| --+-----+ | | | | | | |
| S-NO | No of Convolution Layers | | Dropout Present | | Batch Norm Present | Test Loss |
| Test Acuracy | | | | | | |
| --+-----+-----+-----+----- | | | | | | |
| --+-----+ | | | | | | |
| 1 | | 2 | Present- (Rate=0.25) | | Not present | 0.031 |
| | 0.9897 | | | | | |
| 2 | | 2 | Not present | | Present | 0.023 |
| | 0.9936 | | | | | |
| 3 | | 2 | Present- (Rate=0.5) | | Present | 0.032 |
| | 0.9896 | | | | | |
| 4 | | 3 | Not present | | Present | 0.045 |
| | 0.9874 | | | | | |
| 5 | | 3 | Present- (Rate=0.5) | | Present | 0.023 |
| | 0.9927 | | | | | |
| 6 | | 3 | Not present | | Present | 0.033 |
| | 0.9927 | | | | | |
| 7 | | 5 | Not present | | Present | 0.032 |
| | 0.9913 | | | | | |
| 8 | | 5 | Present- (Rate=0.5) | | Present | 0.038 |
| | 0.99 | | | | | |
| 9 | | 7 | Present- (Rate=0.2) | | Present | 0.025 |
| | 0.9938 | | | | | |
| -----+-----+-----+----- | | | | | | |
| --+-----+ | | | | | | |

In [0]:

```
F='Not present'
T='Present'
T5 = 'Present-(Rate = 0.5)'
T3 = 'Present-(Rate = 0.3)'
T4 = 'Present-(Rate = 0.4)'
Optimiser = ['AdaDelta', 'AdaDelta', 'AdaDelta', 'AdaDelta', 'RMSProp', 'RMSProp', 'RMSProp', 'SGD', 'Adamax']
Activation = ['Relu', 'Tanh', 'Tanh', 'Sigmoid', 'Elu', 'Tanh', 'Tanh', 'Relu', 'Relu']
Initialiser = ['Random_Uniform', 'Random_Uniform', 'Orthogonal', 'Orthogonal', 'Zeros', 'Ones', 'Random_Uniform', 'Random_Uniform', 'Random_Uniform']
Filter_size= [2,2,2,3,3,3,3,3,3]
layers = [3,3,3,3,5,5,5,5,5]
d=[F,F,T4,F,F,F,F,T5,T3]
b=[F,F,T,F,T,T,F,T,F]
test_scr=[0.044,0.063,0.075,2.300,2.301,2.124,2.301,0.193,0.023]
test_acc=[0.986,0.982,0.977,0.113,0.113,0.185,0.097,0.949,0.993]
sno =[1,2,3,4,5,6,7,8,9]
from prettytable import PrettyTable
table = PrettyTable()
table.add_column('S-NO',sno)
table.add_column("No of Convolution Layers",layers)
table.add_column("Optimizer",Optimiser)
table.add_column("Activation fn",Activation)
table.add_column("Initializer",Initialiser)
table.add_column("Size of the filter",Filter_size)
table.add_column("Dropout Present",d)
table.add_column("Batch Norm Present",b)
table.add_column("Test Loss",test_scr)
table.add_column("Test Acuracy",test_acc)
```

In [54]:

```
print('Comparison table of models with different activation functions,intialisers,Optimis
ers other than those used in previous models')
print('In the pretty table Dropout is considered only of the convolutional layers.Some mo
dels are having a dense layer with dropout value 0.5 and some doesnt have any dropout rate
e to their dense layer')
print(table)
```

Comparison table of models with different activation functions,intialisers,Optimisers oth
er than those used in previous models

In the pretty table Dropout is considered only of the convolutional layers.Some models ar
e having a dense layer with dropout value 0.5 and some doesnt have any dropout rate to th

| e having a dense layer with dropout value 0.5 and some doesn't have any dropout rate to their dense layer | | | | | | |
|---|---|--------------------------|--|--------------------|---------------|----------------|
| S-NO | | No of Convolution Layers | | Optimizer | Activation fn | Initializer |
| the filter | | Dropout Present | | Batch Norm Present | Test Loss | Test Accuracy |
| 1 | 3 | Not present | | AdaDelta | Relu | Random_Uniform |
| 2 | 3 | Not present | | AdaDelta | Tanh | Random_Uniform |
| 3 | 3 | Not present | | AdaDelta | Tanh | Orthogonal |
| 4 | 3 | Present-(Rate = 0.4) | | Present | 0.075 | 0.977 |
| 5 | 3 | Not present | | AdaDelta | Sigmoid | Orthogonal |
| 6 | 5 | Not present | | RMSProp | Elu | Zeros |
| 7 | 5 | Not present | | Present | 2.301 | 0.113 |
| 8 | 5 | Not present | | RMSProp | Tanh | Ones |
| 9 | 5 | Not present | | Present | 2.124 | 0.185 |
| 10 | 5 | Not present | | RMSProp | Tanh | Random_Uniform |
| 11 | 5 | Not present | | Not present | 2.301 | 0.097 |
| 12 | 5 | Present-(Rate = 0.5) | | SGD | Relu | Random_Uniform |
| 13 | 5 | Present-(Rate = 0.3) | | Adamax | Relu | Random_Uniform |
| 14 | 5 | Not present | | Not present | 0.023 | 0.993 |

OBSERVATIONS

- In this assignment , I started with loading the MNIST dataset and did some preprocessing on the data.
 - Later , I experimented training different CNN Architectures with Adam Optimiser. I experimented with Batch Normalisation and different values of Dropout.
 - Also , I tried other architectures with different kernal initialisers , activation functions and optimisers.
- During this Assignment , i have observed the following :
- > All the CNN architectures with Adam optimiser achieved high accuracy (above 99 %)
 - > AdaDelta optimiser gave 11 % accuracy with sigmoid activation and achieved 97-99% accuracies when tried with other activation functions
 - > RMSProp optimiser failed to achieve good results, ended up with accuracy of just 10%-20% irrespective of combinations of different activation functions and initialisers.
 - > SGD optimiser model started with low accuracy and then it improved gradually with number of epochs.It may achieve 99% like other good models if trained further for 5-10 epochs.
 - > I believe the show stealer is Adamax optimiser model. It achieved whopping 99.3% test accuracy . It started with low accuracy at the earlier epochs and improved faster than SGD model gradually with number of epochs.