In [1]:

https://arxiv.org/pdf/1608.06993.pdf

from IPython.display import IFrame, YouTubeVideo YouTubeVideo(id = '-W6y8xnd--U', width = 600)

Out[1]:

In [2]:

import tensorflow as tf

The default version of TensorFlow in Colab will soon switch to TensorFlow 2.x.

We recommend you upgrade now or ensure your notebook will continue to use TensorFlow 1.x via the %tensorflow_version 1.x magic: more info.

In [3]:

```
# Load CIFAR-10 Dataset
```

 $(X_train,\ Y_train),\ (X_test,\ Y_test) = tf.keras.datasets.cifar10.load_data()\\ image_height,\ image_width,\ channel = X_train.shape[1],\ X_train.shape[2],\ X_train.shape[3]$

Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz 170500096/170498071 [==========================] - 2s 0us/step

In [4]:

https://stackoverflow.com/a/49961206/10219869

import numpy

print(numpy.unique(Y_train))
print(len(numpy.unique(Y_train).tolist()))

[0 1 2 3 4 5 6 7 8 9]

In [0]:

Y_train to OHE

Y_train = tf.keras.utils.to_categorical(y = Y_train, num_classes = len(numpy.unique(Y_train).tolist()))
Y_test = tf.keras.utils.to_categorical(y = Y_test, num_classes = len(numpy.unique(Y_test).tolist()))

In [6]:

```
print('The shape of X_train is:',X_train.shape) print('The shape of X_test is:',X_test.shape) print('The shape of Y_train is:',Y_train.shape) print('The shape of Y_test is:',Y_test.shape)
```

The shape of X_train is: (50000, 32, 32, 3) The shape of X_test is: (10000, 32, 32, 3) The shape of Y_train is: (50000, 10)

```
The shape of Y_test is: (10000, 10)
```

In [7]:

https://machinelearningmastery.com/image-augmentation-deep-learning-keras/

```
import matplotlib.pyplot as plt
for i in range(0, 9):
  plt.subplot(330 + 1 + i)
  plt.imshow(X_train[i])
plt.show()
```

In [0]:

```
"""ImageDataGenerator(featurewise_center=False, samplewise_center=False, featurewise_std_normalization=False,
            samplewise std normalization=False, zca whitening=False, zca epsilon=1e-06, rotation range=0,
            width_shift_range=0.0, height_shift_range=0.0, brightness_range=None, shear_range=0.0, zoom_range=0.0,
            channel shift range=0.0, fill mode='nearest', cval=0.0, horizontal flip=False, vertical flip=False,
            rescale=None, preprocessing function=None, data format='channels last', validation split=0.0,
            interpolation_order=1, dtype='float32')
batch_size = 128
datagen = tf.keras.preprocessing.image.lmageDataGenerator(featurewise_center= False, samplewise_center= False,
                       featurewise_std_normalization= False, samplewise_std_normalization=False,
                       zca_whitening=False, zca_epsilon=1e-06, rotation_range= 0, width_shift_range=0.1,
                       height_shift_range=0.1, brightness_range=None, shear_range=0.0, zoom_range=0.0,
                       channel_shift_range=0.0, fill_mode='nearest', cval=0.0, horizontal_flip= True,
                       vertical_flip=False, rescale=None, preprocessing_function=None,
                       data_format= None, validation_split=0.0, dtype='float32')
"""datagen test = tf.keras.preprocessing.image.lmageDataGenerator(rescale= 1./255.)"""
datagen.fit(X_train)
train_datagen = datagen.flow(x = X_train, y = Y_train, batch_size = batch_size)
test_datagen = datagen.flow(x = X_test, y = Y_test, batch_size = batch_size)
```

In [0]:

Transition Block

```
compression = 0.5
num_filter = 42

# Dense Block

def dense_block(input, num_filter, dropout_rate = 0):
    global compression
    temp = input

for _ in range(10):
    Batch_Norm = tf.keras.layers.BatchNormalization()(temp)
    Relu = tf.keras.layers.Activation('relu')(Batch_Norm)
    Conv2D = tf.keras.layers.Conv2D(filters= int(num_filter * compression), kernel_size = (3, 3), padding= 'same',)(Relu)
    if dropout_rate > 0:
        Conv2D = tf.keras.layers.Dropout(dropout_rate)(Conv2D)
        Concat = tf.keras.layers.Concatenate(axis= -1)([temp, Conv2D])
        temp = Concat

return temp
```

```
def transition_block(input, num_filter, dropout_rate = 0):
 global compression
 Batch_Norm = tf.keras.layers.BatchNormalization()(input)
 Relu = tf.keras.layers.Activation('relu')(Batch Norm)
 Conv2D_Bottle_Neck = tf.keras.layers.Conv2D(filters= int(num_filter * compression), kernel_size = (1, 1), padding= 'same',)(Relu)
 if dropout_rate > 0:
   Conv2D Bottle Neck = tf.keras.layers.Dropout(dropout rate)(Conv2D Bottle Neck)
 Average pooling = tf.keras.layers.AveragePooling2D(pool_size = (2, 2), strides= 2)(Conv2D_Bottle_Neck)
 return Average pooling
# Output Layer
def output layer(input):
global compression
 Batch_Norm = tf.keras.layers.BatchNormalization()(input)
 Relu = tf.keras.layers.Activation('relu')(Batch_Norm)
 Average pooling = tf.keras.layers.AveragePooling2D(pool size = (2, 2))(Relu)
 Conv = tf.keras.layers.Conv2D(filters= 10, kernel_size = (3, 3), padding= 'same', activation = 'softmax')(Average_pooling)
 Global_Max_pooling = tf.keras.layers.GlobalMaxPooling2D()(Conv)
output = tf.keras.layers.Activation('softmax')(Global Max pooling)
 return output
```

In [0]:

```
Input = tf.keras.layers.Input(shape = (image_height, image_width, channel))
First_Conv2D = tf.keras.layers.Conv2D(filters= num_filter, kernel_size = (3, 3), padding= 'same')(Input)

First_Block = dense_block(First_Conv2D, num_filter, dropout_rate=0)
First_Transition = transition_block(First_Block, num_filter, dropout_rate=0)

Second_Block = dense_block(First_Transition, num_filter, dropout_rate=0)
Second_Transition = transition_block(Second_Block, num_filter, dropout_rate=0)

Third_Block = dense_block(Second_Transition, num_filter, dropout_rate=0)

Third_Transition = transition_block(Third_Block, num_filter, dropout_rate=0)

Last_Block = dense_block(Third_Transition, num_filter, dropout_rate=0)
output = output_layer(input = Last_Block)
```

In [22]:

model = tf.keras.models.Model(inputs = [Input], outputs = [output])
model.summary()

Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to
input_4 (InputLayer)	[(None, 32, 32,	======================================	
conv2d_135 (Conv2D)	(None, 32, 32	2, 42) 1176	input_4[0][0]
batch_normalization_1	32 (BatchN (None, 3	2, 32, 42) 1	68 conv2d_135[0][0]
activation_135 (Activat	ion) (None, 32, 32	42) 0	batch_normalization_132[0][0]
conv2d_136 (Conv2D)	(None, 32, 32	2, 21) 7959	activation_135[0][0]
concatenate_120 (Con		2, 32, 63) 0 nv2d_136[0]	
batch_normalization_1	33 (BatchN (None, 3	2, 32, 63) 2	252 concatenate_120[0][0]
activation_136 (Activat	ion) (None, 32, 32	63) 0	batch_normalization_133[0][0]
conv2d_137 (Conv2D)	(None, 32, 32	2, 21) 1192	8 activation_136[0][0]
concatenate_121 (Con		2, 32, 84) 0 nv2d_137[0]	
batch_normalization_1	34 (BatchN (None, 3	2, 32, 84) 3	336 concatenate_121[0][0]
activation_137 (Activat	ion) (None, 32, 32	84) 0	batch_normalization_134[0][0]

conv2d_138 (Conv2D) (None, 32, 32, 21) 15897 activation_137[0][0]
concatenate_122 (Concatenate) (None, 32, 32, 105) 0 concatenate_121[0][0] conv2d_138[0][0]
batch_normalization_135 (BatchN (None, 32, 32, 105) 420 concatenate_122[0][0]
activation_138 (Activation) (None, 32, 32, 105) 0 batch_normalization_135[0][0]
conv2d_139 (Conv2D) (None, 32, 32, 21) 19866 activation_138[0][0]
concatenate_123 (Concatenate) (None, 32, 32, 126) 0 concatenate_122[0][0] conv2d_139[0][0]
batch_normalization_136 (BatchN (None, 32, 32, 126) 504 concatenate_123[0][0]
activation_139 (Activation) (None, 32, 32, 126) 0 batch_normalization_136[0][0]
conv2d_140 (Conv2D) (None, 32, 32, 21) 23835 activation_139[0][0]
concatenate_124 (Concatenate) (None, 32, 32, 147) 0 concatenate_123[0][0] conv2d_140[0][0]
batch_normalization_137 (BatchN (None, 32, 32, 147) 588 concatenate_124[0][0]
activation_140 (Activation) (None, 32, 32, 147) 0 batch_normalization_137[0][0]
conv2d_141 (Conv2D) (None, 32, 32, 21) 27804 activation_140[0][0]
concatenate_125 (Concatenate) (None, 32, 32, 168) 0 concatenate_124[0][0] conv2d_141[0][0]
batch_normalization_138 (BatchN (None, 32, 32, 168) 672 concatenate_125[0][0]
activation_141 (Activation) (None, 32, 32, 168) 0 batch_normalization_138[0][0]
conv2d_142 (Conv2D) (None, 32, 32, 21) 31773 activation_141[0][0]
concatenate_126 (Concatenate) (None, 32, 32, 189) 0 concatenate_125[0][0] conv2d_142[0][0]
batch_normalization_139 (BatchN (None, 32, 32, 189) 756 concatenate_126[0][0]
activation_142 (Activation) (None, 32, 32, 189) 0 batch_normalization_139[0][0]
conv2d_143 (Conv2D) (None, 32, 32, 21) 35742 activation_142[0][0]
concatenate_127 (Concatenate) (None, 32, 32, 210) 0 concatenate_126[0][0] conv2d_143[0][0]
batch_normalization_140 (BatchN (None, 32, 32, 210) 840 concatenate_127[0][0]
activation_143 (Activation) (None, 32, 32, 210) 0 batch_normalization_140[0][0]
conv2d_144 (Conv2D) (None, 32, 32, 21) 39711 activation_143[0][0]
concatenate_128 (Concatenate) (None, 32, 32, 231) 0 concatenate_127[0][0] conv2d_144[0][0]
batch_normalization_141 (BatchN (None, 32, 32, 231) 924 concatenate_128[0][0]
activation_144 (Activation) (None, 32, 32, 231) 0 batch_normalization_141[0][0]
conv2d_145 (Conv2D) (None, 32, 32, 21) 43680 activation_144[0][0]
concatenate_129 (Concatenate) (None, 32, 32, 252) 0 concatenate_128[0][0] conv2d_145[0][0]
batch_normalization_142 (BatchN (None, 32, 32, 252) 1008 concatenate_129[0][0]
activation_145 (Activation) (None, 32, 32, 252) 0 batch_normalization_142[0][0]
conv2d_146 (Conv2D) (None, 32, 32, 21) 5313 activation_145[0][0]
average_pooling2d_12 (AveragePo (None, 16, 16, 21) 0 conv2d_146[0][0]
batch_normalization_143 (BatchN (None, 16, 16, 21) 84 average_pooling2d_12[0][0]
activation_146 (Activation) (None, 16, 16, 21) 0 batch_normalization_143[0][0]
conv2d_147 (Conv2D) (None, 16, 16, 21) 3990 activation_146[0][0]

concatenate_130 (Concatenate) (None, 16, 16, 42) 0 average_poolingzd_12[0][0] conv2d_147[0][0]
batch_normalization_144 (BatchN (None, 16, 16, 42) 168 concatenate_130[0][0]
activation_147 (Activation) (None, 16, 16, 42) 0 batch_normalization_144[0][0]
conv2d_148 (Conv2D) (None, 16, 16, 21) 7959 activation_147[0][0]
concatenate_131 (Concatenate) (None, 16, 16, 63) 0 concatenate_130[0][0] conv2d_148[0][0]
batch_normalization_145 (BatchN (None, 16, 16, 63) 252 concatenate_131[0][0]
activation_148 (Activation) (None, 16, 16, 63) 0 batch_normalization_145[0][0]
conv2d_149 (Conv2D) (None, 16, 16, 21) 11928 activation_148[0][0]
concatenate_132 (Concatenate) (None, 16, 16, 84) 0
batch_normalization_146 (BatchN (None, 16, 16, 84) 336 concatenate_132[0][0]
activation_149 (Activation) (None, 16, 16, 84) 0 batch_normalization_146[0][0]
conv2d_150 (Conv2D) (None, 16, 16, 21) 15897 activation_149[0][0]
concatenate_133 (Concatenate) (None, 16, 16, 105) 0 concatenate_132[0][0] conv2d_150[0][0]
batch_normalization_147 (BatchN (None, 16, 16, 105) 420 concatenate_133[0][0]
activation_150 (Activation) (None, 16, 16, 105) 0 batch_normalization_147[0][0]
conv2d_151 (Conv2D) (None, 16, 16, 21) 19866 activation_150[0][0]
concatenate_134 (Concatenate) (None, 16, 16, 126) 0
batch_normalization_148 (BatchN (None, 16, 16, 126) 504 concatenate_134[0][0]
activation_151 (Activation) (None, 16, 16, 126) 0 batch_normalization_148[0][0]
conv2d_152 (Conv2D) (None, 16, 16, 21) 23835 activation_151[0][0]
concatenate_135 (Concatenate) (None, 16, 16, 147) 0 concatenate_134[0][0] conv2d_152[0][0]
batch_normalization_149 (BatchN (None, 16, 16, 147) 588 concatenate_135[0][0]
activation_152 (Activation) (None, 16, 16, 147) 0 batch_normalization_149[0][0]
conv2d_153 (Conv2D) (None, 16, 16, 21) 27804 activation_152[0][0]
concatenate_136 (Concatenate) (None, 16, 16, 168) 0 concatenate_135[0][0] conv2d_153[0][0]
batch_normalization_150 (BatchN (None, 16, 16, 168) 672 concatenate_136[0][0]
activation_153 (Activation) (None, 16, 16, 168) 0 batch_normalization_150[0][0]
conv2d_154 (Conv2D) (None, 16, 16, 21) 31773 activation_153[0][0]
concatenate_137 (Concatenate) (None, 16, 16, 189) 0 concatenate_136[0][0] conv2d_154[0][0]
batch_normalization_151 (BatchN (None, 16, 16, 189) 756 concatenate_137[0][0]
activation_154 (Activation) (None, 16, 16, 189) 0 batch_normalization_151[0][0]
conv2d_155 (Conv2D) (None, 16, 16, 21) 35742 activation_154[0][0]
concatenate_138 (Concatenate) (None, 16, 16, 210) 0 concatenate_137[0][0] conv2d_155[0][0]
batch_normalization_152 (BatchN (None, 16, 16, 210) 840 concatenate_138[0][0]
activation_155 (Activation) (None, 16, 16, 210) 0 batch_normalization_152[0][0]
conv2d_156 (Conv2D) (None, 16, 16, 21) 39711 activation_155[0][0]
concatenate_139 (Concatenate) (None, 16, 16, 231) 0 concatenate_138[0][0] conv2d 156[0][0]

batch_normalization_153 (BatchN (None, 16, 16, 231) 924 concatenate_139[0][0]
activation_156 (Activation) (None, 16, 16, 231) 0 batch_normalization_153[0][0]
conv2d_157 (Conv2D) (None, 16, 16, 21) 4872 activation_156[0][0]
average_pooling2d_13 (AveragePo (None, 8, 8, 21) 0 conv2d_157[0][0]
batch_normalization_154 (BatchN (None, 8, 8, 21) 84 average_pooling2d_13[0][0]
activation_157 (Activation) (None, 8, 8, 21) 0 batch_normalization_154[0][0]
conv2d_158 (Conv2D) (None, 8, 8, 21) 3990 activation_157[0][0]
concatenate_140 (Concatenate) (None, 8, 8, 42) 0 average_pooling2d_13[0][0] conv2d_158[0][0]
batch_normalization_155 (BatchN (None, 8, 8, 42) 168 concatenate_140[0][0]
activation_158 (Activation) (None, 8, 8, 42) 0 batch_normalization_155[0][0]
conv2d_159 (Conv2D) (None, 8, 8, 21) 7959 activation_158[0][0]
concatenate_141 (Concatenate) (None, 8, 8, 63) 0 concatenate_140[0][0] conv2d_159[0][0]
batch_normalization_156 (BatchN (None, 8, 8, 63) 252 concatenate_141[0][0]
activation_159 (Activation) (None, 8, 8, 63) 0 batch_normalization_156[0][0]
conv2d_160 (Conv2D) (None, 8, 8, 21) 11928 activation_159[0][0]
concatenate_142 (Concatenate) (None, 8, 8, 84) 0 concatenate_141[0][0] conv2d_160[0][0]
batch_normalization_157 (BatchN (None, 8, 8, 84) 336 concatenate_142[0][0]
activation_160 (Activation) (None, 8, 8, 84) 0 batch_normalization_157[0][0]
conv2d_161 (Conv2D) (None, 8, 8, 21) 15897 activation_160[0][0]
concatenate_143 (Concatenate) (None, 8, 8, 105) 0 concatenate_142[0][0] conv2d_161[0][0]
batch_normalization_158 (BatchN (None, 8, 8, 105) 420 concatenate_143[0][0]
activation_161 (Activation) (None, 8, 8, 105) 0 batch_normalization_158[0][0]
conv2d_162 (Conv2D) (None, 8, 8, 21) 19866 activation_161[0][0]
concatenate_144 (Concatenate) (None, 8, 8, 126) 0 concatenate_143[0][0] conv2d_162[0][0]
batch_normalization_159 (BatchN (None, 8, 8, 126) 504 concatenate_144[0][0]
activation_162 (Activation) (None, 8, 8, 126) 0 batch_normalization_159[0][0]
conv2d_163 (Conv2D) (None, 8, 8, 21) 23835 activation_162[0][0]
concatenate_145 (Concatenate) (None, 8, 8, 147) 0 concatenate_144[0][0] conv2d_163[0][0]
batch_normalization_160 (BatchN (None, 8, 8, 147) 588 concatenate_145[0][0]
activation_163 (Activation) (None, 8, 8, 147) 0 batch_normalization_160[0][0]
conv2d_164 (Conv2D) (None, 8, 8, 21) 27804 activation_163[0][0]
concatenate_146 (Concatenate) (None, 8, 8, 168) 0 concatenate_145[0][0] conv2d_164[0][0]
batch_normalization_161 (BatchN (None, 8, 8, 168) 672 concatenate_146[0][0]
activation_164 (Activation) (None, 8, 8, 168) 0 batch_normalization_161[0][0]
conv2d_165 (Conv2D) (None, 8, 8, 21) 31773 activation_164[0][0]
concatenate_147 (Concatenate) (None, 8, 8, 189) 0
batch_normalization_162 (BatchN (None, 8, 8, 189) 756 concatenate_147[0][0]

activation_165 (Activation) (None, 8, 8, 189) 0 batch_normalization_162[0][0]
conv2d_166 (Conv2D) (None, 8, 8, 21) 35742 activation_165[0][0]
concatenate_148 (Concatenate) (None, 8, 8, 210) 0 concatenate_147[0][0] conv2d_166[0][0]
batch_normalization_163 (BatchN (None, 8, 8, 210) 840 concatenate_148[0][0]
activation_166 (Activation) (None, 8, 8, 210) 0 batch_normalization_163[0][0]
conv2d_167 (Conv2D) (None, 8, 8, 21) 39711 activation_166[0][0]
concatenate_149 (Concatenate) (None, 8, 8, 231) 0 concatenate_148[0][0] conv2d_167[0][0]
batch_normalization_164 (BatchN (None, 8, 8, 231) 924 concatenate_149[0][0]
activation_167 (Activation) (None, 8, 8, 231) 0 batch_normalization_164[0][0]
conv2d_168 (Conv2D) (None, 8, 8, 21) 4872 activation_167[0][0]
average_pooling2d_14 (AveragePo (None, 4, 4, 21) 0 conv2d_168[0][0]
batch_normalization_165 (BatchN (None, 4, 4, 21) 84 average_pooling2d_14[0][0]
activation_168 (Activation) (None, 4, 4, 21) 0 batch_normalization_165[0][0]
conv2d_169 (Conv2D) (None, 4, 4, 21) 3990 activation_168[0][0]
concatenate_150 (Concatenate) (None, 4, 4, 42) 0 average_pooling2d_14[0][0] conv2d_169[0][0]
batch_normalization_166 (BatchN (None, 4, 4, 42) 168 concatenate_150[0][0]
activation_169 (Activation) (None, 4, 4, 42) 0 batch_normalization_166[0][0]
conv2d_170 (Conv2D) (None, 4, 4, 21) 7959 activation_169[0][0]
concatenate_151 (Concatenate) (None, 4, 4, 63) 0 concatenate_150[0][0] conv2d_170[0][0]
batch_normalization_167 (BatchN (None, 4, 4, 63) 252 concatenate_151[0][0]
activation_170 (Activation) (None, 4, 4, 63) 0 batch_normalization_167[0][0]
conv2d_171 (Conv2D) (None, 4, 4, 21) 11928 activation_170[0][0]
concatenate_152 (Concatenate) (None, 4, 4, 84) 0 concatenate_151[0][0] conv2d_171[0][0]
batch_normalization_168 (BatchN (None, 4, 4, 84) 336 concatenate_152[0][0]
activation_171 (Activation) (None, 4, 4, 84) 0 batch_normalization_168[0][0]
conv2d_172 (Conv2D) (None, 4, 4, 21) 15897 activation_171[0][0]
concatenate_153 (Concatenate) (None, 4, 4, 105) 0 concatenate_152[0][0] conv2d_172[0][0]
batch_normalization_169 (BatchN (None, 4, 4, 105) 420 concatenate_153[0][0]
activation_172 (Activation) (None, 4, 4, 105) 0 batch_normalization_169[0][0]
conv2d_173 (Conv2D) (None, 4, 4, 21) 19866 activation_172[0][0]
concatenate_154 (Concatenate) (None, 4, 4, 126) 0 concatenate_153[0][0] conv2d_173[0][0]
batch_normalization_170 (BatchN (None, 4, 4, 126) 504 concatenate_154[0][0]
activation_173 (Activation) (None, 4, 4, 126) 0 batch_normalization_170[0][0]
conv2d_174 (Conv2D) (None, 4, 4, 21) 23835 activation_173[0][0]
concatenate_155 (Concatenate) (None, 4, 4, 147) 0 concatenate_154[0][0] conv2d_174[0][0]
batch_normalization_171 (BatchN (None, 4, 4, 147) 588 concatenate_155[0][0]
activation_174 (Activation) (None, 4, 4, 147) 0 batch_normalization_171[0][0]

```
conv2d_175 (Conv2D)
                          (None, 4, 4, 21) 27804
                                                    activation_174[0][0]
concatenate_156 (Concatenate) (None, 4, 4, 168) 0
                                                     concatenate_155[0][0]
                                   conv2d_175[0][0]
batch_normalization_172 (BatchN (None, 4, 4, 168) 672
                                                       concatenate_156[0][0]
activation_175 (Activation)
                         (None, 4, 4, 168) 0
                                                 batch_normalization_172[0][0]
conv2d_176 (Conv2D)
                          (None, 4, 4, 21)
                                          31773
                                                    activation_175[0][0]
concatenate 157 (Concatenate) (None, 4, 4, 189) 0
                                                     concatenate 156[0][0]
                                   conv2d_176[0][0]
batch_normalization_173 (BatchN (None, 4, 4, 189) 756
                                                       concatenate_157[0][0]
activation_176 (Activation)
                         (None, 4, 4, 189) 0
                                                 batch_normalization_173[0][0]
conv2d_177 (Conv2D)
                          (None, 4, 4, 21) 35742
                                                    activation_176[0][0]
concatenate_158 (Concatenate) (None, 4, 4, 210) 0
                                                     concatenate_157[0][0]
                                   conv2d_177[0][0]
batch_normalization_174 (BatchN (None, 4, 4, 210) 840
                                                       concatenate_158[0][0]
activation_177 (Activation)
                         (None, 4, 4, 210) 0
                                                 batch_normalization_174[0][0]
conv2d 178 (Conv2D)
                          (None, 4, 4, 21) 39711
                                                    activation_177[0][0]
concatenate_159 (Concatenate) (None, 4, 4, 231) 0
                                                     concatenate_158[0][0]
                                   conv2d_178[0][0]
batch_normalization_175 (BatchN (None, 4, 4, 231) 924
                                                       concatenate_159[0][0]
activation_178 (Activation) (None, 4, 4, 231) 0
                                                 batch_normalization_175[0][0]
average_pooling2d_15 (AveragePo (None, 2, 2, 231) 0
                                                       activation_178[0][0]
conv2d_179 (Conv2D)
                          (None, 2, 2, 10)
                                          20800
                                                    average_pooling2d_15[0][0]
global_max_pooling2d_2 (GlobalM (None, 10)
                                              0
                                                     conv2d_179[0][0]
activation_179 (Activation) (None, 10)
                                        0
                                                global_max_pooling2d_2[0][0]
______
Total params: 973,843
Trainable params: 962,293
Non-trainable params: 11,550
```

In [0]:

```
# determine loss function and optimizer
```

model.compile(loss = 'categorical_crossentropy', optimizer= tf.keras.optimizers.Adam(learning_rate = 0.001), metrics=['accuracy'],)

In [24]:

```
# Resource Exaustion Error, https://github.com/tensorflow/models/issues/1993
batch_size = 128
history = model.fit_generator(generator = train_datagen, steps_per_epoch = numpy.floor(len(X_train) / batch_size), epochs= 200,
   validation data = test datagen
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
   389/390 [====
Epoch 6/200
390/390 [===
```

```
Epoch 7/200
Epoch 8/200
390/390 [============] - 63s 161ms/step - loss: 1.6531 - acc: 0.7197 - val_loss: 1.7481 - val_acc: 0.6132
Epoch 9/200
390/390 [==========] - 63s 161ms/step - loss: 1.6420 - acc: 0.7339 - val_loss: 1.6667 - val_acc: 0.6935
Epoch 10/200
390/390 [============] - 63s 161ms/step - loss: 1.6333 - acc: 0.7470 - val loss: 1.6868 - val acc: 0.6762
Epoch 11/200
390/390 [=============] - 63s 161ms/step - loss: 1.6275 - acc: 0.7564 - val loss: 1.7352 - val acc: 0.6433
390/390 [============] - 63s 161ms/step - loss: 1.6230 - acc: 0.7655 - val_loss: 1.6951 - val_acc: 0.6732
Epoch 13/200
390/390 [============] - 62s 160ms/step - loss: 1.6155 - acc: 0.7747 - val_loss: 1.6424 - val_acc: 0.7489
Epoch 14/200
Epoch 15/200
390/390 [============] - 62s 159ms/step - loss: 1.6058 - acc: 0.7966 - val loss: 1.6660 - val acc: 0.7272
Epoch 16/200
390/390 [===========] - 62s 160ms/step - loss: 1.5998 - acc: 0.8098 - val_loss: 1.6340 - val_acc: 0.7710
Epoch 17/200
390/390 [============] - 63s 161ms/step - loss: 1.5953 - acc: 0.8129 - val_loss: 1.6480 - val_acc: 0.7513
Epoch 18/200
Epoch 19/200
390/390 [============] - 63s 161ms/step - loss: 1.5902 - acc: 0.8209 - val_loss: 1.6336 - val_acc: 0.7739
Epoch 20/200
390/390 [============] - 62s 160ms/step - loss: 1.5874 - acc: 0.8237 - val loss: 1.6338 - val acc: 0.7686
Epoch 21/200
390/390 [===========] - 62s 160ms/step - loss: 1.5834 - acc: 0.8299 - val_loss: 1.6321 - val_acc: 0.7675
Epoch 22/200
390/390 [=============] - 63s 161ms/step - loss: 1.5845 - acc: 0.8307 - val_loss: 1.6300 - val_acc: 0.7839
Epoch 24/200
Epoch 25/200
Epoch 26/200
390/390 [============] - 63s 161ms/step - loss: 1.5721 - acc: 0.8459 - val_loss: 1.6181 - val_acc: 0.7890
Epoch 27/200
390/390 [=============] - 63s 160ms/step - loss: 1.5703 - acc: 0.8479 - val_loss: 1.6823 - val_acc: 0.7179
Epoch 28/200
390/390 [==========] - 62s 160ms/step - loss: 1.5651 - acc: 0.8553 - val_loss: 1.6267 - val_acc: 0.7636
Epoch 29/200
Epoch 30/200
390/390 [============] - 62s 160ms/step - loss: 1.5639 - acc: 0.8582 - val_loss: 1.6263 - val_acc: 0.7735
Epoch 31/200
390/390 [===========] - 63s 161ms/step - loss: 1.5597 - acc: 0.8638 - val_loss: 1.6428 - val_acc: 0.7626
Epoch 32/200
390/390 [===========] - 63s 161ms/step - loss: 1.5591 - acc: 0.8648 - val_loss: 1.6167 - val_acc: 0.7997
Epoch 33/200
Epoch 34/200
```

```
390/390 [====
Epoch 35/200
Epoch 36/200
390/390 [============] - 63s 160ms/step - loss: 1.5538 - acc: 0.8727 - val loss: 1.5865 - val acc: 0.8285
Epoch 37/200
Epoch 38/200
Epoch 39/200
390/390 [=============] - 63s 160ms/step - loss: 1.5501 - acc: 0.8758 - val loss: 1.6256 - val acc: 0.7912
Epoch 40/200
390/390 [===========] - 63s 161ms/step - loss: 1.5473 - acc: 0.8806 - val_loss: 1.5973 - val_acc: 0.8154
Epoch 41/200
390/390 [============] - 63s 161ms/step - loss: 1.5464 - acc: 0.8814 - val_loss: 1.5870 - val_acc: 0.8380
Epoch 42/200
390/390 [============] - 63s 162ms/step - loss: 1.5437 - acc: 0.8844 - val_loss: 1.5768 - val_acc: 0.8452
Epoch 43/200
390/390 [============] - 63s 162ms/step - loss: 1.5432 - acc: 0.8857 - val loss: 1.5806 - val acc: 0.8416
Epoch 44/200
390/390 [============] - 63s 161ms/step - loss: 1.5409 - acc: 0.8881 - val loss: 1.6459 - val acc: 0.7655
Epoch 45/200
Epoch 46/200
390/390 [=============] - 63s 163ms/step - loss: 1.5401 - acc: 0.8897 - val loss: 1.5884 - val acc: 0.8296
Epoch 47/200
Epoch 48/200
390/390 [============] - 63s 162ms/step - loss: 1.5378 - acc: 0.8920 - val loss: 1.6062 - val acc: 0.8102
Epoch 49/200
390/390 [============] - 63s 162ms/step - loss: 1.5369 - acc: 0.8936 - val loss: 1.5633 - val acc: 0.8607
Epoch 50/200
Epoch 51/200
390/390 [============] - 63s 162ms/step - loss: 1.5350 - acc: 0.8964 - val_loss: 1.6052 - val_acc: 0.8074
Epoch 52/200
390/390 [=========] - 63s 162ms/step - loss: 1.5317 - acc: 0.9001 - val_loss: 1.6030 - val_acc: 0.8069
Epoch 53/200
390/390 [==========] - 63s 163ms/step - loss: 1.5324 - acc: 0.9004 - val_loss: 1.5957 - val_acc: 0.8229
Epoch 54/200
390/390 [============] - 63s 162ms/step - loss: 1.5303 - acc: 0.9006 - val loss: 1.5730 - val acc: 0.8469
Epoch 55/200
390/390 [============] - 63s 162ms/step - loss: 1.5309 - acc: 0.9017 - val loss: 1.5834 - val acc: 0.8345
Epoch 56/200
390/390 [=============] - 63s 162ms/step - loss: 1.5298 - acc: 0.9047 - val loss: 1.5730 - val acc: 0.8485
Epoch 57/200
390/390 [============] - 63s 162ms/step - loss: 1.5288 - acc: 0.9049 - val_loss: 1.5806 - val_acc: 0.8414
Epoch 58/200
Epoch 59/200
Epoch 60/200
Epoch 61/200
```

```
Epoch 62/200
390/390 [============] - 63s 162ms/step - loss: 1.5244 - acc: 0.9100 - val_loss: 1.5790 - val_acc: 0.8436
Epoch 64/200
390/390 [==========] - 63s 162ms/step - loss: 1.5238 - acc: 0.9129 - val_loss: 1.5760 - val_acc: 0.8418
Epoch 65/200
390/390 [===========] - 63s 163ms/step - loss: 1.5231 - acc: 0.9134 - val_loss: 1.5728 - val_acc: 0.8481
Epoch 66/200
390/390 [============] - 63s 162ms/step - loss: 1.5233 - acc: 0.9112 - val loss: 1.5927 - val acc: 0.8280
Epoch 67/200
390/390 [=============] - 63s 163ms/step - loss: 1.5207 - acc: 0.9158 - val loss: 1.5564 - val acc: 0.8702
Epoch 68/200
Epoch 69/200
390/390 [=============] - 63s 162ms/step - loss: 1.5200 - acc: 0.9166 - val_loss: 1.5595 - val_acc: 0.8653
Epoch 70/200
390/390 [============] - 63s 162ms/step - loss: 1.5196 - acc: 0.9185 - val loss: 1.5705 - val acc: 0.8576
Epoch 71/200
Epoch 72/200
390/390 [============] - 63s 162ms/step - loss: 1.5182 - acc: 0.9185 - val_loss: 1.5569 - val_acc: 0.8704
Epoch 73/200
390/390 [==========] - 63s 163ms/step - loss: 1.5160 - acc: 0.9217 - val_loss: 1.5638 - val_acc: 0.8673
Epoch 74/200
390/390 [==========] - 63s 162ms/step - loss: 1.5162 - acc: 0.9214 - val_loss: 1.5868 - val_acc: 0.8395
Epoch 75/200
390/390 [=============] - 63s 162ms/step - loss: 1.5157 - acc: 0.9235 - val loss: 1.5624 - val acc: 0.8644
Epoch 76/200
390/390 [============] - 63s 162ms/step - loss: 1.5146 - acc: 0.9223 - val loss: 1.5599 - val acc: 0.8658
Epoch 77/200
390/390 [============] - 63s 162ms/step - loss: 1.5161 - acc: 0.9236 - val loss: 1.5559 - val acc: 0.8710
Epoch 78/200
Epoch 79/200
390/390 [=============] - 63s 162ms/step - loss: 1.5129 - acc: 0.9268 - val_loss: 1.5724 - val_acc: 0.8537
Epoch 80/200
390/390 [===========] - 63s 163ms/step - loss: 1.5130 - acc: 0.9276 - val_loss: 1.5685 - val_acc: 0.8599
Epoch 81/200
390/390 [=============] - 63s 162ms/step - loss: 1.5140 - acc: 0.9256 - val loss: 1.5862 - val acc: 0.8413
Epoch 82/200
390/390 [=============] - 63s 162ms/step - loss: 1.5133 - acc: 0.9262 - val_loss: 1.5721 - val_acc: 0.8575
Epoch 83/200
390/390 [===========] - 63s 162ms/step - loss: 1.5122 - acc: 0.9276 - val_loss: 1.5570 - val_acc: 0.8671
Epoch 84/200
Epoch 85/200
Epoch 86/200
390/390 [============] - 63s 162ms/step - loss: 1.5076 - acc: 0.9321 - val loss: 1.5661 - val acc: 0.8642
Epoch 87/200
Epoch 88/200
390/390 [===========] - 63s 161ms/step - loss: 1.5097 - acc: 0.9300 - val_loss: 1.5564 - val_acc: 0.8745
Epoch 89/200
```

```
389/390 1===
         =========>.| - ETA: US - IOSS: 1.5084 - acc: 0.9341Epoch 1/200
390/390 [==============] - 63s 161ms/step - loss: 1.5084 - acc: 0.9340 - val_loss: 1.5605 - val_acc: 0.8664
Epoch 90/200
390/390 [===========] - 63s 161ms/step - loss: 1.5092 - acc: 0.9314 - val_loss: 1.5577 - val_acc: 0.8680
Epoch 91/200
Epoch 92/200
Epoch 93/200
390/390 [============] - 63s 160ms/step - loss: 1.5058 - acc: 0.9366 - val_loss: 1.5587 - val_acc: 0.8692
Epoch 94/200
390/390 [==========] - 63s 161ms/step - loss: 1.5063 - acc: 0.9345 - val_loss: 1.5621 - val_acc: 0.8685
Epoch 95/200
390/390 [=============] - 63s 161ms/step - loss: 1.5067 - acc: 0.9359 - val loss: 1.5499 - val acc: 0.8830
Epoch 96/200
390/390 [=============] - 63s 160ms/step - loss: 1.5073 - acc: 0.9355 - val_loss: 1.5738 - val_acc: 0.8553
Epoch 97/200
Epoch 98/200
Epoch 99/200
390/390 [==============] - 62s 160ms/step - loss: 1.5031 - acc: 0.9398 - val_loss: 1.5560 - val_acc: 0.8705
Epoch 100/200
390/390 [===========] - 63s 161ms/step - loss: 1.5048 - acc: 0.9379 - val_loss: 1.5526 - val_acc: 0.8783
Epoch 101/200
390/390 [===========] - 63s 160ms/step - loss: 1.5049 - acc: 0.9374 - val_loss: 1.5589 - val_acc: 0.8683
Epoch 102/200
Epoch 103/200
Epoch 104/200
390/390 [============] - 63s 161ms/step - loss: 1.5032 - acc: 0.9398 - val loss: 1.5623 - val acc: 0.8718
Epoch 105/200
Epoch 106/200
Epoch 107/200
Epoch 108/200
Epoch 109/200
390/390 [=================] - 62s 160ms/step - loss: 1.5012 - acc: 0.9443 - val loss: 1.5572 - val acc: 0.8772
Epoch 110/200
390/390 [=============] - 62s 160ms/step - loss: 1.5003 - acc: 0.9445 - val loss: 1.5614 - val acc: 0.8717
Epoch 111/200
390/390 [============] - 62s 160ms/step - loss: 1.5009 - acc: 0.9448 - val_loss: 1.5662 - val_acc: 0.8628
Epoch 112/200
390/390 [===========] - 62s 160ms/step - loss: 1.4994 - acc: 0.9456 - val_loss: 1.5567 - val_acc: 0.8717
Epoch 113/200
Epoch 114/200
390/390 [=============] - 62s 159ms/step - loss: 1.5001 - acc: 0.9442 - val_loss: 1.5514 - val_acc: 0.8799
Epoch 115/200
390/390 [============] - 62s 160ms/step - loss: 1.4979 - acc: 0.9476 - val_loss: 1.5558 - val_acc: 0.8769
Epoch 116/200
390/390 [=============] - 62s 160ms/step - loss: 1.4986 - acc: 0.9468 - val loss: 1.5498 - val acc: 0.8866
```

```
Epoch 117/200
390/390 [===========] - 62s 160ms/step - loss: 1.4994 - acc: 0.9448 - val_loss: 1.5538 - val_acc: 0.8767
Epoch 119/200
390/390 [=====
Epoch 120/200
Epoch 121/200
390/390 [=============] - 63s 161ms/step - loss: 1.4957 - acc: 0.9503 - val_loss: 1.5483 - val_acc: 0.8900
Epoch 122/200
390/390 [============] - 63s 161ms/step - loss: 1.4964 - acc: 0.9501 - val_loss: 1.5478 - val_acc: 0.8886
Epoch 123/200
390/390 [=============] - 62s 160ms/step - loss: 1.4945 - acc: 0.9529 - val loss: 1.5610 - val acc: 0.8708
Froch 124/200
390/390 [============] - 63s 160ms/step - loss: 1.4960 - acc: 0.9497 - val_loss: 1.5538 - val_acc: 0.8750
Epoch 125/200
390/390 [============] - 62s 159ms/step - loss: 1.4962 - acc: 0.9501 - val_loss: 1.5424 - val_acc: 0.8939
Epoch 126/200
390/390 [=============] - 62s 159ms/step - loss: 1.4934 - acc: 0.9546 - val_loss: 1.5594 - val_acc: 0.8753
Epoch 127/200
390/390 [===========] - 62s 158ms/step - loss: 1.4961 - acc: 0.9509 - val_loss: 1.5419 - val_acc: 0.8934
Epoch 128/200
Epoch 129/200
Epoch 130/200
Epoch 131/200
Epoch 132/200
390/390 [===========] - 62s 159ms/step - loss: 1.4943 - acc: 0.9527 - val_loss: 1.5513 - val_acc: 0.8835
Epoch 133/200
390/390 [===========] - 62s 159ms/step - loss: 1.4943 - acc: 0.9538 - val_loss: 1.5613 - val_acc: 0.8671
Epoch 134/200
390/390 [=============] - 62s 159ms/step - loss: 1.4941 - acc: 0.9529 - val loss: 1.5476 - val acc: 0.8845
Epoch 135/200
390/390 [=============] - 62s 159ms/step - loss: 1.4937 - acc: 0.9535 - val_loss: 1.5546 - val_acc: 0.8767
Epoch 136/200
390/390 [=========] - 63s 160ms/step - loss: 1.4939 - acc: 0.9531 - val_loss: 1.5475 - val_acc: 0.8853
Epoch 137/200
Epoch 138/200
390/390 [============] - 62s 160ms/step - loss: 1.4929 - acc: 0.9542 - val loss: 1.5667 - val acc: 0.8578
Epoch 139/200
Epoch 140/200
Epoch 141/200
390/390 [============] - 63s 161ms/step - loss: 1.4932 - acc: 0.9548 - val_loss: 1.5486 - val_acc: 0.8834
Epoch 142/200
Epoch 143/200
390/390 [============] - 62s 159ms/step - loss: 1.4923 - acc: 0.9562 - val_loss: 1.5502 - val_acc: 0.8841
Epoch 144/200
```

```
305/35U I==
             ========>.| - ETA. 05 - 1055. 1.4910 - acc. 0.9073Epocit 1/200
390/390 [==========] - 62s 158ms/step - loss: 1.4918 - acc: 0.9574 - val_loss: 1.5442 - val_acc: 0.8853
Epoch 145/200
389/390 [===
      Epoch 146/200
Epoch 147/200
Epoch 148/200
390/390 [============] - 62s 158ms/step - loss: 1.4902 - acc: 0.9587 - val_loss: 1.5540 - val_acc: 0.8794
Epoch 149/200
390/390 [============] - 62s 159ms/step - loss: 1.4895 - acc: 0.9598 - val loss: 1.5831 - val acc: 0.8403
390/390 [=============] - 62s 158ms/step - loss: 1.4899 - acc: 0.9580 - val loss: 1.5440 - val acc: 0.8918
Epoch 151/200
390/390 [============] - 62s 159ms/step - loss: 1.4893 - acc: 0.9590 - val_loss: 1.5509 - val_acc: 0.8849
Epoch 152/200
390/390 [===========] - 62s 158ms/step - loss: 1.4882 - acc: 0.9613 - val_loss: 1.5528 - val_acc: 0.8837
Epoch 153/200
390/390 [===========] - 62s 158ms/step - loss: 1.4893 - acc: 0.9592 - val_loss: 1.5514 - val_acc: 0.8833
Epoch 154/200
Epoch 155/200
389/390 [======
       390/390 [=============] - 62s 158ms/step - loss: 1.4896 - acc: 0.9601 - val loss: 1.5421 - val acc: 0.8944
Epoch 156/200
390/390 [==========] - 62s 159ms/step - loss: 1.4881 - acc: 0.9606 - val_loss: 1.5711 - val_acc: 0.8623
Epoch 157/200
Epoch 158/200
390/390 [=============] - 62s 159ms/step - loss: 1.4873 - acc: 0.9635 - val_loss: 1.5427 - val_acc: 0.8959
Epoch 159/200
390/390 [==========] - 62s 158ms/step - loss: 1.4880 - acc: 0.9618 - val_loss: 1.5428 - val_acc: 0.8920
Epoch 160/200
390/390 [============] - 62s 159ms/step - loss: 1.4878 - acc: 0.9620 - val_loss: 1.5451 - val_acc: 0.8910
390/390 [============] - 62s 158ms/step - loss: 1.4881 - acc: 0.9609 - val loss: 1.5560 - val acc: 0.8792
Epoch 162/200
Epoch 163/200
390/390 [===========] - 62s 159ms/step - loss: 1.4869 - acc: 0.9621 - val_loss: 1.5501 - val_acc: 0.8852
Epoch 164/200
390/390 [==========] - 62s 159ms/step - loss: 1.4863 - acc: 0.9649 - val_loss: 1.5435 - val_acc: 0.8930
Epoch 165/200
390/390 [============] - 62s 159ms/step - loss: 1.4881 - acc: 0.9613 - val loss: 1.5461 - val acc: 0.8903
Epoch 166/200
390/390 [==========] - 62s 158ms/step - loss: 1.4885 - acc: 0.9610 - val_loss: 1.5418 - val_acc: 0.8926
Epoch 167/200
390/390 [=============] - 61s 158ms/step - loss: 1.4868 - acc: 0.9644 - val_loss: 1.5514 - val_acc: 0.8812
Epoch 168/200
390/390 [==========] - 62s 158ms/step - loss: 1.4875 - acc: 0.9631 - val_loss: 1.5535 - val_acc: 0.8760
Epoch 169/200
390/390 [=============] - 62s 158ms/step - loss: 1.4852 - acc: 0.9655 - val_loss: 1.5386 - val_acc: 0.9013
Epoch 170/200
390/390 [=============] - 62s 159ms/step - loss: 1.4850 - acc: 0.9665 - val loss: 1.5566 - val acc: 0.8818
Epoch 171/200
```

```
Epoch 172/200
390/390 [=============] - 62s 159ms/step - loss: 1.4873 - acc: 0.9627 - val loss: 1.5508 - val acc: 0.8867
Epoch 174/200
Epoch 175/200
Epoch 176/200
Epoch 177/200
390/390 [============] - 62s 159ms/step - loss: 1.4844 - acc: 0.9664 - val_loss: 1.5443 - val_acc: 0.8938
Epoch 178/200
390/390 [==========] - 62s 158ms/step - loss: 1.4848 - acc: 0.9665 - val_loss: 1.5413 - val_acc: 0.8928
Epoch 179/200
390/390 [============] - 62s 160ms/step - loss: 1.4838 - acc: 0.9672 - val loss: 1.5441 - val acc: 0.8943
Epoch 180/200
390/390 [=============] - 62s 159ms/step - loss: 1.4850 - acc: 0.9670 - val loss: 1.5401 - val acc: 0.8992
Epoch 181/200
390/390 [=============] - 62s 159ms/step - loss: 1.4855 - acc: 0.9660 - val_loss: 1.5437 - val_acc: 0.8947
Epoch 182/200
Epoch 183/200
390/390 [============] - 62s 160ms/step - loss: 1.4841 - acc: 0.9676 - val loss: 1.5368 - val acc: 0.9034
Epoch 184/200
390/390 [=============] - 62s 160ms/step - loss: 1.4844 - acc: 0.9677 - val loss: 1.5508 - val acc: 0.8846
Epoch 185/200
390/390 [===========] - 62s 160ms/step - loss: 1.4844 - acc: 0.9661 - val_loss: 1.5499 - val_acc: 0.8895
Epoch 186/200
Epoch 187/200
Epoch 188/200
Epoch 189/200
390/390 [=============] - 62s 159ms/step - loss: 1.4823 - acc: 0.9699 - val loss: 1.5519 - val acc: 0.8858
Epoch 190/200
390/390 [============] - 62s 159ms/step - loss: 1.4830 - acc: 0.9691 - val loss: 1.5493 - val acc: 0.8846
Epoch 191/200
390/390 [============] - 62s 159ms/step - loss: 1.4839 - acc: 0.9683 - val_loss: 1.5430 - val_acc: 0.8982
Epoch 192/200
Epoch 193/200
Epoch 194/200
390/390 [============] - 62s 160ms/step - loss: 1.4823 - acc: 0.9703 - val loss: 1.5447 - val acc: 0.8914
Epoch 195/200
390/390 [============] - 62s 159ms/step - loss: 1.4838 - acc: 0.9676 - val_loss: 1.5498 - val_acc: 0.8824
Epoch 196/200
390/390 [============] - 62s 160ms/step - loss: 1.4813 - acc: 0.9705 - val_loss: 1.5456 - val_acc: 0.8893
Epoch 197/200
Epoch 198/200
Epoch 199/200
```

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In [0]:

```
# We save the model

model.save('model_dense')
```

In [28]:

```
import matplotlib.pyplot as plt
score = model.evaluate(x= X_test, y= Y_test)
print('Test Loss:', score[0])
print()
print('Test Accuracy:', score[1])
def plt_dynamic(x, val_y, test_y, ax, color='b'):
 ax.plot(x, val_y, 'b', label = 'Validation_loss')
 ax.plot(x, test_y, 'r', label = 'Test_loss')
 plt.grid()
 plt.legend()
 fig.canvas.draw()
fig, ax = plt.subplots(1, 1, figsize = (12, 8))
ax.set_xlabel('Epoch')
ax.set_ylabel('Categorical Cross Entropy')
plt.title('Our Own Dense Network')
x = list(range(1, 201))
val_y = history.history['val_loss']
test_y = history.history['loss']
plt_dynamic(x, val_y, test_y, ax)
```

10000/10000 [==============] - 5s 500us/sample - loss: 1.5389 - acc: 0.9023

Test Loss: 1.5389170993804933

Test Accuracy: 0.9023

