Maggie Cao
23245277
1738 63rd Street 2nd FL
Brooklyn, NY 11204
(212) 380-3142
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Paula Whitlock
Brooklyn College
Project Interim Status Report

Below are important research of neural networks which will be implemented in the code:

Steps to create a neural network:

- 1. Learn a model that generates sensory data rather than classifying it. Eliminates the need for large amounts of labeled data.
- 2. Learn one layer of representation at a time using restricted boltzmann machines. This decomposes the overall learning task into multiple simpler tasks and eliminates the inference problems that arise in generative models.
- 3. Use a separate fine-tuning stage to improve the generative or discriminative abilities of the composite model.

A combination of these ideas leads to a novel and effective way of learning multiple layers of representation.

- Geoffrey E. Hinton

Optimization:

Steps to improve on a neural network from Geoffrey E. Hinton:

Allow higher-level feature detectors to communicate their needs to lower-level ones whilst also being easy to implement in layered networks of stochastic binary neurons that have activation states of 1 or 0 turned on with a probability that is a smooth non-linear function of the total input they receive.

Without the layer-by-layer learning, fine-tuning alone is hopelessly slow. Instead of fine-tuning the model to be a better at generating data, back-propagation can be used to fine-tune it to be better at discrimination. This works well.

To infer a probability distribution over the various possible settings of the hidden variables.

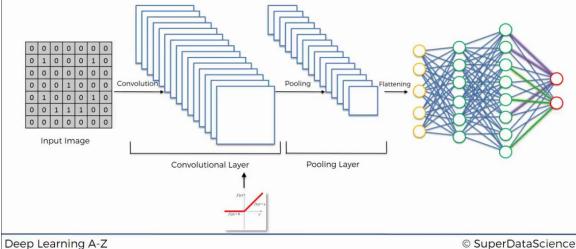
Gaussian distribution, Restricted Boltzmann Machines.

Learning feature detectors

The optimizer function in Kera's classifier.compile(optimizer, loss, metrics) is the algorithm you are going to use to find the optimal set of weights of the network. The "adam" optimizer using stochastic gradient descent algorithm that's efficient. What about the rmsprop? It computes the single gradient in batches and is slower. A sigmoid loss function is similar to logistic regression. After weight updates, the model uses metrics accuracy to improve the model's performance.

CNN Architecture:

2DConv -> ReLU -> MaxPool -> 2DConv -> ReLU -> MaxPool -> Flatten() -> Fully connected 2-layer neural network



128 neurons for the first layer -> ReLU -> 128 for hidden laver -> ReLU -> 3 neurons for output layer -> softmax

Learning/Training

The training process will use the cross-entropy error with activation functions of sigmoid or softmax. The softmax produces probability of the output. The starting loss, given at training, need to be consistent with the number of classes in the network. The training process will use stochastic gradient where the gradient is computed per input instead of in a batch. I will also try rmsprop, which is a batch training. I also forgot to use the prediction function if the output is 0/1 but that can be adjusted for a multi-class output. Here's an example from the "Deep-Learning in Python" on-line lecture that uses a simple ANN:

#Part 3: Making predictions and evaluating the model #Predicting the test results

y_prediction = classifier.predict(x_test_scaled)

 $y_prediction = (y_prediction > 0.5)$

#neural network's final output will be true if the activation function is greater than 0.5, which means greater than 50% chance of leaving the bank

#Predicting a single new observation

new_prediction = classifier.predict(sc.transform(np.array([[0.0,0,600,1,40,3,60000,2,1,1,50000]])))

new_prediction = (new_prediction > 0.5)

#Making the Confusion Matrix

from sklearn.metrics import confusion_matrix

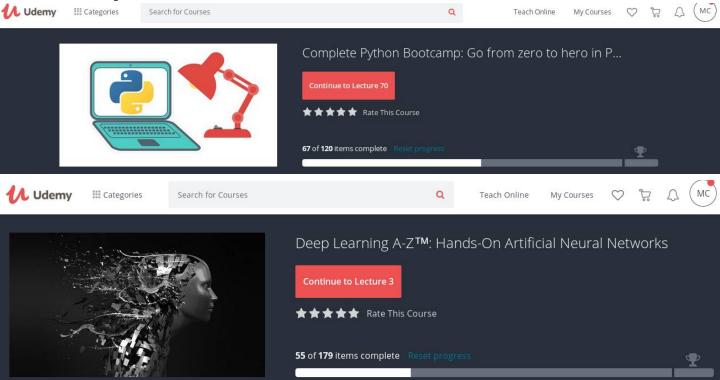
cm = confusion_matrix(y_test, y_pred) #so far we just split your dataset into a training set and a test set The variance problem of using validation sets is because validation sets can represent very different accuracy on another test, which is very inconsistent. Judging model on just one accuracy and one test set is not super relevant for knowing how well the model does in terms of loss, accuracy and generalization. The K-Fold Cross Validation will fix this variance problem because it splits the training set into 10 folds where k = 10 in 10 different iterations. Nine folds will represent the training set and 1 fold is to test the neural network. It is much more relevant because it takes the average.

First few weeks of September:

Research on Neural Network's and programming in Python

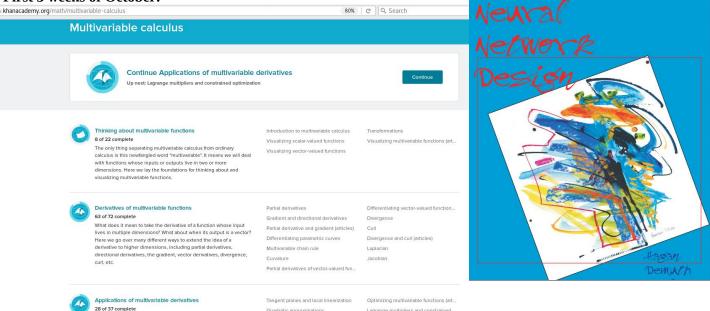
Paid \$100 to go to an in-person group for deep learning, which uses the cloud to train on images of cats and dogs. The lecturer told me I should use Tensorflow or one of the popular libraries. Since I'm interested in extracting features of shapes for the neural network to learn, he told me that a convolutional network will do the job. This is because a convolutional neural network is designed to learn the pixels of images in a three dimensional output space. It does this by pooling and flattening the layers of a constant pixel size, or use padding if the size doesn't fit the dimensions of the image.

Last 3 weeks of September:



I spent this time taking udemy's online courses in learning the basics of python, first two week's of Andrew Ng's machine learning course. I have tried training a basic convolutional neural network of cats and dogs using the tutorial online but since my laptop doesn't have a Nvidia GPU I can't use GPU computation locally. It will take a couple of days just to get the output of the convolutional network.

First 3 weeks of October:



I decided to use the machine learning library Keras instead because it uses Tensorflow (in python 3) and Theano (in python 2) as backend. I spent 3 weeks reading Hagan's Neural Network Design book (2 weeks), reviewing on linear algebra (1 week) and learning and taking notes on multi-variable calculus on Kahn academy (1 week).

Week of October 23:

The baby AI image dataset is very old and has bugs in it. I wasn't able to extract the dataset by running their python program. So, I spent all this time creating my own dataset and preparing it for loading using pickle's serialization format into Google Cloud's Machine Learning Engine. I created my own python class called Draw.py, which uses

multiprocessing of Pool workers in a class to draw images themselves as well as the intersection of images. Multiprocessing allows me to make as many images as possible by using parallel computing of 4 cores in a CPU.

```
canvas.save img(filename)
47 def draw square(object, filename):
         canvas = object(500,500)
bg_obj = canvas.background_color()
49
         canvas.line_square(bg_obj)
50
51
52
53
54
         context_obj = canvas.new_context()
canvas.fill_square(context_obj)
         canvas.save_img(filename)
55 def draw_triangle(object, filename):
56 canvas = object(500,500)
57 bg_obj = canvas.background_color()
         canvas.line_triangle(bg_obj)
         context_obj = canvas.new_context()
canvas.fill_triangle(context_obj)
59
60
         canvas.save_img(filename)
    class Draw(object):
                init (self, canvas width, canvas height):
               import numpy as np
self.canvas width = canvas width
66
67
68
69
70
71
72
73
74
75
76
77
78
80
81
82
               self.canvas_height = canvas_height
self.data = np.zeros ((self.canvas_width, self.canvas_height, 4),
               self.surface = cairo.ImageSurface.create_for_data(self.data,
                                                                                      cairo.FORMAT_ARGB32,
                                                                                      self.canvas_width,
self.canvas_height)
         def run(self):
               p = Pool(processes=4)
              p.apply_async(draw_triangle, (Draw,str(x)) )
p.close()
               p.join()
         def new context(self):
               return cairo.Context(self.surface)
```

```
draw.py ⋈
   Created on Fri Oct 27 18:41:03 2017
 3 Draws images of shapes of circles, rectangles, squares, triangles
 4 @author: maggie
          future__ import print_function
 7 import cairo
 8 import random
 9 from multiprocessing import Pool
11 def draw_objects(object, filename):
       canvas = object(500,500)
       obil = canvas.background color()
       canvas.fill_circle(obj1)
       obj2 = canvas.new context()
16
17
18
       canvas.line_circle(obj2)
       obj3 = canvas.new context()
       canvas.line triangle(obj3)
       obj4 = canvas.new_context()
20
21
22
23
       canvas.fill_triangle(obj4)
       obj5 = canvas.new context()
       canvas.line rectangle(obj5)
       obj6 = canvas.new_context()
24
25
26
27
28
       canvas.fill rectangle(obj6)
       obj7 = canvas.new_context()
       canvas.line square(obi7)
       obj8 = canvas.new context()
       canvas.fill_square(obj8)
29
30
       canvas.save_img(filename)
31 def draw rectangle(object, filename):
       canvas = object(500,500)
       bg_obj = canvas.background_color()
       canvas.line_rectangle(bg_obj)
35
       context_obj = canvas.new_context()
canvas.fill_rectangle(context_obj)
       canvas.save_img(filename)
39 def draw circle(object, filename):
       canvas = object(500,500)
       bg_obj = canvas.background_color()
       canvas.fill_circle(bg_obj)
       context_obj = canvas.new_context()
       canvas.line circle(context_obj)
```

```
90
                 r = random.uniform(0,1)
                                                                                    133
                                                                                                   object.set_source_rgb(r, g, b)
                 g = random.uniform(0,1)
b = random.uniform(0,1)
  91
                                                                                    134
                                                                                                   object.fill()
  92
                                                                                    135
                 name = self.new_context()
  93
                                                                                    136
                                                                                              def line_rectangle(self, object):
  94
                 name.set_source_rgb(r,g,b)
                                                                                    137
                                                                                                    r = random.uniform(0,1)
  95
                 name.paint()
                                                                                    138
                                                                                                    g = random.uniform(0,1)
  96
                 return name
                                                                                    139
                                                                                                   b = random.uniform(0,1)
  97
                                                                                    140
                                                                                                   x = random.randint(10,500)
           def fill_circle(self, object):
  98
                                                                                    141
                                                                                                   y = random.randint(10,500)
  99
                 import math
                                                                                    142
                                                                                                   w = random.uniform(0,10)
                 r = random.uniform(0,1)
  100
                                                                                    143
                                                                                                   width = random.randint(50,250)
  101
                 g = random.uniform(0,1)
                                                                                    144
                                                                                                   height = random.randint(50,250)
                 \tilde{b} = random.uniform(0,1)
  102
                                                                                                   object.rectangle(x, y, width, height)
object.set_line_width(w)
                                                                                    145
  103
                 xc = random.randint(10,500)
                                                                                    146
                 yc = random.randint(10,500)
  104
                                                                                    147
                                                                                                   object.set_source_rgb(r, g, b)
                 radius = random.randint(50,250)
  105
                                                                                    148
                                                                                                   object.stroke()
  106
                 object.arc(xc, yc, radius, 0, 2*math.pi)
                                                                                    149
                 object.set_source_rgb(r, g, b)
object.fill()
  107
                                                                                    150
                                                                                              def fill_square(self, object):
  108
                                                                                    151
                                                                                                   r = random.uniform(0,1)
  109
                                                                                    152
                                                                                                   g = random.uniform(0,1)
           def line_circle(self, object):
  110
                                                                                    153
                                                                                                   b = random.uniform(0,1)
                 import math
                                                                                    154
                                                                                                   x = random.randint(10,500)
                 r = random.uniform(0,1)
                                                                                    155
                                                                                                    y = random.randint(10,500)
                 g = random.uniform(0,1)
                                                                                    156
                                                                                                    width = random.randint(50,250)
  114
                 b = random.uniform(0.1)
                                                                                    157
                                                                                                   object.rectangle(x, y, width, width)
  115
                 xc = random.randint(10,500)
                                                                                    158
                                                                                                   object.set_source_rgb(r, g, b)
                 yc = random.randint(10,500)
                                                                                    159
                                                                                                   object.fill()
  117
                 radius = random.randint(50,250)
  118
                 w = random.uniform(0,10)
                                                                                              def line_square(self, object):
                                                                                    161
  119
                 object.arc(xc, yc, radius, 0, 2*math.pi)
                                                                                                   r = random.uniform(0,1)
                                                                                    162
                                                                                                   g = random.uniform(0,1)
  120
                 object.set_line_width(w)
                                                                                    163
                 object.set_source_rgb(r, g, b)
                                                                                    164
                                                                                                   b = random.uniform(0,1)
                 object.stroke()
                                                                                                   x = random.randint(10,500)
                                                                                    165
  123
                                                                                                   y = random.randint(10,500)
                                                                                    166
                                                                                                    w = random.uniform(0,10)
  124
           def fill rectangle(self, object):
                                                                                    167
                r = random.uniform(0,1)
                                                                                                   width = random.randint(50,250)
  125
                                                                                                   object.rectangle(x, y, width, width)
object.set_line_width(w)
  126
                 q = random.uniform(0,1)
                                                                                    169
  127
                 b = random.uniform(0,1)
 128
                x = random.randint(10,500)
171
               object.set_source_rgb(r, g, b)
                                                                                                    This file reduces the image's quality to reduce the file size:
172
               object.stroke()
173
174
         def fill_triangle(self, object):
                                                                                                      2 Created on Thu Oct 26 20:50:49 2017
               r = random.uniform(0,1)
175
176
               g = random.uniform(0,1)
                                                                                                      4 @author: maggie
               b = random.uniform(0.1)
              x = random.randint(10,500)
                                                                                                      6 from __future__ import print_function
7 from future import division
179
              v = random.randint(10.500)
                                                                                                      8 from PIL import Image
180
               x1 = random.randint(10,500)
                                                                                                     9 import glob
10 import pickle
11 import scipy.misc
12 import numpy as np
13 from multiprocessing import Lock
14 from multiprocessing import Pool
181
              y1 = random.randint(10,500)
               y2 = random.randint(10,500)
              object.move_to(x,y)
object.line_to(x, y1)
object.line_to(x1, y2)
              object.line_to(x, y)
object.set_source_rgb(r, g, b)
186
                                                                                                     16 def init(lock):
17 global childs_lock
18 childs_lock = lock
187
               object.fill()
         def line_triangle(self, object):
    r = random.uniform(0,1)
    g = random.uniform(0,1)
    b = random.uniform(0,1)
                                                                                                    19 """each pool worker gets original img data to reduce file size"""
21 def reduce images(image_path):
    childs_lock.acquire()
23 img = Image.open(image_path)
190
192
193
                                                                                                           img = image.pointimage_path;
childs lock.release()
basewidth = 300
percent = (basewidth / float(img.size[0]))
hsize = int((float(img.size[1]) * float(percent)))
img = img.resize((basewidth, hsize), Image.ANTIALIAS) #ANTIALIAS reserves quality
* train = np.array(img)
194
              x = random.randint(10,500)
              y = random.randint(10,500)
               x1 = random.randint(10,500)
               y1 = random.randint(10,500)
                                                                                                            x_train = np.array(img)
               y2 = random.randint(10,500)
                                                                                                              train = np.array(img, dtype = np.uint8) #a numpy array with data type CV 8UC1 train = x_train[:,:,0] #slice out the color dimension
                                                                                                     30
31
               w = random.uniform(0,3)
199
              object.move_to(x,y)
object.line_to(x, y1)
200
                                                                                                            print (x_train.shape)
201
                                                                                                            img.close()
              object.line_to(x1, y2)
object.line_to(x, y)
object.set_line_width(w)
object.set_source_rgb(r, g, b)
202
                                                                                                            return x_train
203
                                                                                                     36 #global storage variable for both main and pool of workers
204
                                                                                                     37 result_list = []
205
               object.stroke()
206
                                                                                                     39 """result(data) is called whenever process_images(path) returns a result
207
                                                                                                    40 result_list is modified by main process not by pool of workers" 41 def result(data):
         def save_img(self, filename):
208
              print (filename)
dir = "test_set/triangle/"
                                                                                                           result_list.append(data)
209
               intersection = "triangle."
212
               self.surface.write_to_png(dir + intersection + filename + ".png")
214 if
           name == ' main
          d = Draw(500, 500)
215
216
          d.run()
217
```

y = random.randint(10,500)

width = random.randint(50,250)

height = random.randint(50,250)

object.rectangle(x, y, width, height)

130

131

132

#the next image drawn on background color must have

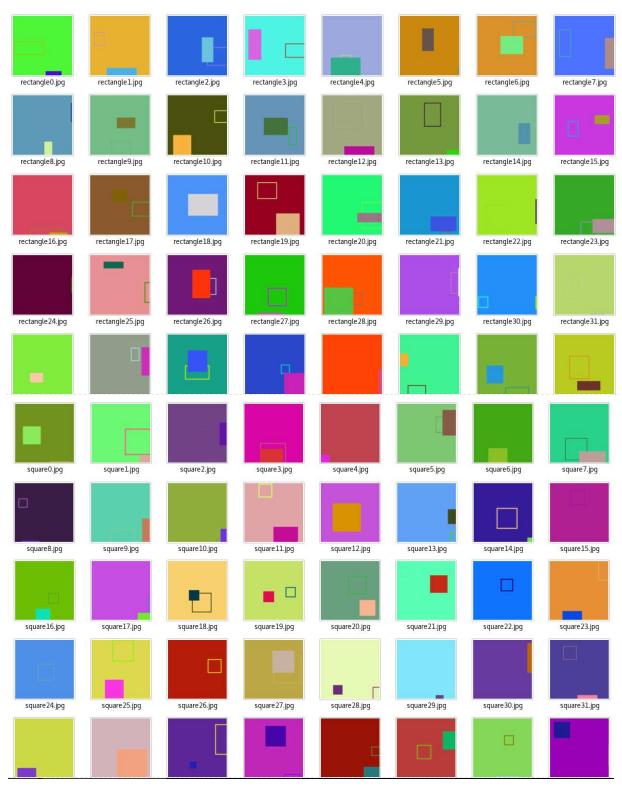
#the parameter in background color

def background color(self):

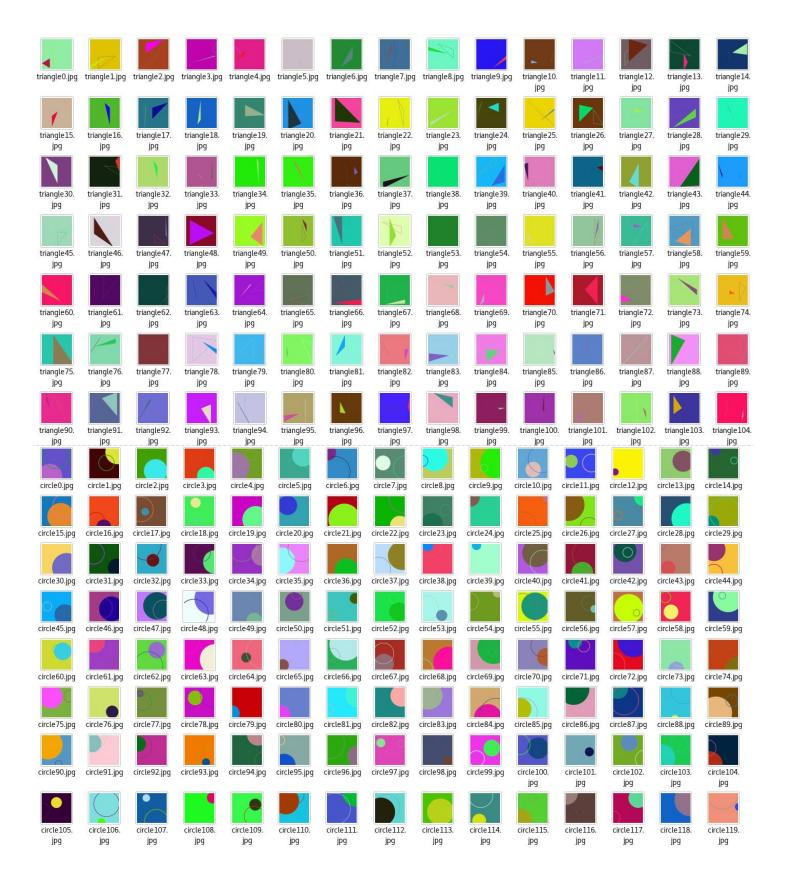
88

89

```
44 #create empty pickle file first then append to file 45 output = open (pickle_file, 'wb')
46 output.close()
47
48 def result(data):
          output = open (pickle_file, 'ab')
print ("in pickle file: " , pickle_file)
49
50
51
          pickle.dump(data, output, pickle.HIGHEST_PROTOCOL)
52
          output.close()
53
54 if __name__ == '__main__':
55
56
          shape_path = "test_set/circle1/"
57
          lock = Lock()
58
          p = Pool(processes=4, initargs = (lock, ), initializer = init)
59
          for image_path in glob.glob(shape_path + "*jpg"):
60
61
                p.apply_async(process_images, (image_path, shape_path), callback = result)
62
          p.close() # no more tasks
          p.join() #wrap up current tasks
63
64
     circle0.jpg
                          circle1.jpg
                                              circle2.jpg
                                                                   circle3.jpg
                                                                                        circle4.jpg
                                                                                                             circle5.jpg
                                                                                                                                  circle6.jpg
                                                                                                                                                       circle7.jpg
                                                                   circle11.jpg
                         circle9.jpg
                                              circle10.jpg
                                                                                                             circle13.jpg
                                                                                                                                 circle 14.jpg
                                                                                                                                                      circle15.jpg
                                                                                        circle12.jpg
     circle8.jpg
    circle16.jpg
                                              circle 18.jpg
                         circle17.jpg
                                                                   circle 19.jpg
                                                                                        circle 20.jpg
                                                                                                             circle 21.jpg
                                                                                                                                 circle 22.jpg
                                                                                                                                                      circle23.jpg
    circle 24.jpg
                         circle 25.jpg
                                              circle 26.jpg
                                                                   circle 27.jpg
                                                                                        circle 28.jpg
                                                                                                             circle 29.jpg
                                                                                                                                 circle 30.jpg
                                                                                                                                                      circle31.jpg
    triangle0.jpg
                         triangle1.jpg
                                                                  triangle3.jpg
                                                                                       triangle4.jpg
                                                                                                            triangle5.jpg
                                                                                                                                 triangle6.jpg
                                              triangle 2.jpg
                                                                                                                                                      triangle7.jpg
    triangle8.jpg
                         triangle9.jpg
                                             triangle 10.jpg
                                                                  triangle 11.jpg
                                                                                                                                                      triangle15.jpg
                                                                                       triangle12.jpg
                                                                                                            triangle 13.jpg
                                                                                                                                 triangle 14.jpg
                                                                                                            triangle 21.jpg
                                                                                                                                 triangle 22.jpg
    triangle 16. jpg
                        triangle 17. jpg
                                             triangle 18. jpg
                                                                  triangle 19. jpg
                                                                                       triangle 20. jpg
                                                                                                                                                      triangle 23.jpg
    triangle 24.jpg
                        triangle 25. jpg
                                              triangle 26.jpg
                                                                  triangle 27.jpg
                                                                                       triangle 28.jpg
                                                                                                            triangle 29. jpg
                                                                                                                                 triangle 30. jpg
                                                                                                                                                      triangle 31.jpg
```



Below are more examples of the training dataset in JPEG extension:





I had problems loading the images to a pickle file because I originally stored the images as a dictionary which represents in a string format. Numpy wants a float object, so I decided to use Python's list data structure to store all the numpy arrays.

10.31.17:

The training set consists of a total of 6,200 images. Before being serialized into a pickle file, the training set is organized in a tuple structure (numpy array, y_label). The numpy array is the data array processed by the PIL module in (300, 300, 3) format. The numpy array represents the matrix in float32 of the image. The y_label represents the target values of the shapes, which is the expected output of the convolutional neural network. Keras requires categorical crossentropy loss to be computed with categorical encodings. The categorical one hot encoding transfers integers (0...number of classes) into binary format. My y_label is a series of categorical hot encodings of 0, 1, 2 in binary format of three classes (circles, rectangles and squares, triangle).

I had to change the numpy array data structure from a default float to float32 bit since the loading of the pickle files in the default float structure consumes too much memory in megabytes per file. The difference almost reduced the entire file size from 3.0 GB (without compression) to 1.7 G.B. The pickle files are too huge, so I have to reduce the quality and size of each image to reduce the pickle files. Pickle loads and image creation of the shapes are created using multiprocessing of independent Pool workers. I have been trying to figure out how to create a pickle file, organize numpy arrays and store them in a huge list, dump that huge list using joblib. Use memmap to store large numpy arrays because it's inefficient for the list to increase in data memory allocation in list comprehension of pickle loading. The file below create (numpy arrays, y_label) tuples and stores them in a pickle file.

The short-term goal is to train the shapes individually first and then figure out how to get the model to generalize on the "intersection" of shapes either by using recurrent convolutional neural networks or multi-label output using supervised learning. How will the network learn? I need to adjust the architecture of the CNN. The multi-label output is simpler and much easier. This requires sigmoid activation and loss = binary_crossentropy at the output layer for multi-label output to work.

```
load_merge_files.py > 45 if __name__ == '__main__':
  draw.py 🗶
                googlecloud_config_cnn.txt X
                                                                  46
1 from
                                                                  47
                                                                         circle path = "test set/rectangle/"
         future import print function
 2 from PIL import Image
                                                                  48
                                                                         lock = Lock()
 3 import glob
                                                                  49
                                                                         p = Pool(processes=4, initargs = (lock, ), initializer = init)
 4 import pickle
                                                                  50
 5 import numpy as np
                                                                         for image_path in glob.glob(circle_path + "*png"):
                                                                  51
6 from multiprocessing import Lock
                                                                  52
                                                                             p.apply async(reduce images, (image path,), callback = result)
7 from multiprocessing import Pool
                                                                  53
8 from keras.utils import to_categorical
                                                                  54
                                                                  55
                                                                         p.close() # no more tasks
10 """to make lock and queue storage global to all child workers
                                                                  56
                                                                        p.join() #wrap up current tasks
11 def init(lock):
                                                                  57
      global childs_lock
                                                                  58
                                                                         output = open ('test_rectangle.pkl', 'wb')
13
      childs_lock = lock
                                                                  59
                                                                         for x in result list:
14
                                                                  60
                                                                             pickle.dump(x, output, -1)
15 def process_images(image path, shape path):
                                                                  61
                                                                        output.close()
          shape y = None
16
                                                                  62
          if shape_path == "test_set/circle1/":
                                                                  63
                                                                         name = []
              shape_y = 0
                                                                  64
                                                                         num files = 2000
          elif shape_path == "test_set/rectangle1/":
19
                                                                  65
                                                                         for i in range(num_files):
20
               shape y = 1
          elif shape_path == "test_set/triangle1/":
                                                                             name.append("test_set/rectangle1/rectangle" + str(i) + ".jpg")
                                                                  66
21
22
                                                                  67
              shape_y = 2
23
       elif shape_path == "test_set/square1/":
                                                                         #save resized data to a folder
                                                                  68
                                                                        with open('test rectangle.pkl', 'rb') as pkl file:
24
              shape y = 3
                                                                  69
25
                                                                             data1 = [pickle.load(pkl_file) for i in range(num_files)]
26
          ylabel = to categorical(shape y, num classes = 4)
                                                                  71
                                                                         for i in range(num_files):
27
          ylabel = np.reshape(ylabel, (4))
                                                                             scipy.misc.imsave(name[i], datal[i])
          print ("new shape", ylabel.shape)
28
29
          print (ylabel)
30
          childs_lock.acquire()
31
          img = Image.open(image_path)
32
          childs lock.release()
33
34
          np_img = np.array(img, dtype = [('img_info', np.float16)])
35
             'y_label = np.empty(None, dtype = [('y_class',np.int8)])
          y label.fill(shape_y)'''
37
          # Add another dimension 1 image number for keras to process
          #np_img = np_img.reshape( (-1, ) + np_img.shape)
38
          img.close()
39
          return np_img['img_info'], ylabel
40
42 #global storage variable for both main and pool of workers
43 pickle file = 'test circle.pkl'
```

This file merges all the pickled files that each represents the individual shape data and their y_labels from training, validation and testing set.

```
draw.py X googlecloud_config_cnn.txt X load_merge_files.py X
   1 from __future__ import print_function
     import pickle
   ∃ import joblib
   4 #import numpy as np
5 #from tempfile import mkdtemp
   6 #import os.path as path
   8 def load_train_or_test(files):
9  with open(files, 'rb') as f:
               try:
                    print ("Opening files")
                    print (files)
  13
                    while True:
               yield pickle.load(f) #python version 2.7
except EOFError:
                    pass
  18 if __name__ == '__main__':
          #include (shape array v labels) as a tuple returned by the pickl
          circle_dataset = [item for item in load_train_or_test ("circle.pkl")]
          triangle dataset = [item for item in load_train_or_test ("triangle.pkl")]
rectangle_dataset = [item for item in load_train_or_test ("rectangle.pkl")]
square_dataset = [item for item in load_train_or_test ("square.pkl")]
  25
26
          train_data = circle_dataset + triangle_dataset + rectangle_dataset + square_dataset
  28
29
30
31
32
33
34
35
          validation circle dataset = [item for item in load train or test ("validate circle.pkl")]
          validation_triangle_dataset = [item for item in load_train_or_test ("validate_triangle.pkl")]
          validation_rectangle_dataset = [item for item in load_train_or_test ("validate_rectangle.pkl")]
validation_square_dataset = [item for item in load_train_or_test ("validate_square.pkl")]
          validation_data = validation_circle_dataset + validation_triangle_dataset + validation_rectangle_dataset + validation_square_dataset
  36
37
38
          pickle_file = 'shape_data.pkl'
                                                                                                    f.close()
          f = open(pickle_file, 'wb')
                                                                                              except Exception as e:
                                                                                     46
                                                                                     47
                                                                                                    print('Unable to save data to', pickle_file, ':', e)
  39
                          #'train_shape_dataset': train_shape_dataset,
'train_data': train_data,
               save = {#
                                                                                     48
  40
                                                                                     49
                          validation data': validation data,
                                                                                     50
                                                                                     51
                #pickle.dump(save, f, pickle.HIGHEST PROTOCOL)
  43
               joblib.dump(save, f, compress = True)
                                                                                     52
```

This file uses memory mapping to store large numpy arrays, and randomize the data arrays. It then stores all the data in a compressed pickle file for Google Cloud to load. Google Cloud uses python 2, so the CNN loader file will also use python 2.

```
draw.py 🗶 googlecloud_config_cnn.txt 🗶
                                                           load merge files.py X
  1 from __future__ import print_function
   2 import joblib
  3 import pickle
  4 import numpy as np
  5 from tempfile import mkdtemp
  6 import os.path as path
    '''returns individual list data info and y label data in numpy arrays'''
  9 def get_data(shape_temp_file, label_temp_file, dataset):
temp_filename = path.join(mkdtemp(), shape_temp_file)
         train_shape_dataset = np.memmap(temp_filename, dtype = np.float16, mode = 'w+', shape = (300, 300, 3))
temp_filename1 = path.join(mkdtemp(), label_temp_file)
train_y_dataset = np.memmap(temp_filename1, dtype = np.float16, mode = 'w+', shape = (3))
         train_shape_dataset = [x[0] for x in dataset]
         train_shape_dataset = np.array(train_shape_dataset)
print ("in get_data function for dataset")
         print (train_shape_dataset.shape)
         train_y_dataset = [x[1] for x in dataset]
         train_y_dataset = np.array(train_y_dataset)
         print (train_y_dataset.shape)
         return train shape dataset, train y dataset
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
         pickle_file = 'shape_data.pkl'
         np.random.seed(135)
         with open(pickle_file, 'rb') as f:
              save = joblib.load(f)
              train_data = save['train_data']
              validation_data = save['validation_data']
del save # hint to help gc free up memory
         #shuffle the tuple (shape info, y label) dataset
         np.random.seed(135)
         np.random.shuffle(train_data)
         #split list in half
         train_data_half = train_data[ :: 3]
 45
          validation data half = validation data[ :: 3]
 46
          train_shape_dataset, train_y_dataset = get_data('shapes.dat', 'shapes_y.dat', train_data_half)
 47
 49
          validate shape dataset, validate y dataset = get data('validate shapes.dat', 'validate shapes y.dat', validation da
 50
51
52
          print ("in main: 1/half train", train_shape_dataset.shape)
print ("in main: 1/half y_label", train_y_dataset.shape)
#print ("in main: 2/half train", train_shape_halfdataset.shape)
#print ("in main: 2/half y_label", train_y_halfdataset.shape)
print ("in main: validate", validate_shape_dataset.shape)
print ("in main: validate y_label", validate_y_dataset.shape)
 54
55
56
 57
58
          pickle file = 'random shapes.pkl'
 59
                f = open(pickle_file, 'wb')
 60
 61
                save = {'train_shape_dataset': train_shape_dataset,
 62
                           'train_y_dataset': train_y_dataset,
                           #'train_shape_halfdataset': train_shape_halfdataset,
#'train_y_halfdataset': train_y_halfdataset,
 63
                           'validate shape dataset': validate shape dataset,
                           'validate_y_dataset': validate_y_dataset,
                          }
                #pickle.dump(save,
                                         f, pickle.HIGHEST PROTOCOL)
                joblib.dump(save, f, compress = True)
                f.close()
          except Exception as e:
                print('Unable to save data to', pickle file, ':', e)
                raise
```

11.3-11.5.17:

Google cloud works locally but had errors of loading pickle file remotely on google cloud because the Cloud Compute Engine doesn't recognize python's file descriptor. I need to use tensorflow's open method, need to set gs:// for every input file data for Google Cloud to recognized it. Here are the steps to run the CNN loader file in Google Cloud:

```
draw.py ×
                googlecloud config cnn.txt X
 3 gsutil cp -r trainer/cloudml-gpu.yaml gs://cnninput_dataset/trainer/cloudml-gpu.yaml
  gsutil cp -r trainer/__init__.py gs://cnninput_dataset/trainer/__init_
  gsutil cp -r data/random_shapes.pkl gs://cnninput_dataset/data/random_shapes.pkl
10 gsutil cp -r setup.py gs://cnninput_dataset/setup.py
13 export BUCKET NAME=cnninput dataset
14 export JOB_NAME="cnncopy_train_$(date +%Y%m%d_%H%M%S)"
15 export JOB DIR=qs://$BUCKET NAME/$JOB NAME
16 export REGION=us-east1
18 train on machine locally
19 gcloud ml-engine local train \
    --job-dir $JOB DIR \
    --module-name trainer.cnncopy \
    --package-path ./trainer \
24
    --train-file ./data/random shapes.pkl
26 submit a job to cloud ML engine
27 gcloud ml-engine jobs submit training $JOB NAME \
      --job-dir $JOB DIR \
      --runtime-version 1.0 \
      --module-name trainer.cnncopy \
31
      --package-path ./trainer \
      --region $REGION \
32
33
      --config trainer/cloudml-gpu.yaml \
34
      --train-file gs://$BUCKET_NAME/data/random_shapes.pkl
36
37 submit a job to cloud ML engine
38 gcloud ml-engine jobs submit training $JOB_NAME \
      --job-dir $JOB DIR \
       --runtime-version 1.0 \
41
      --module-name trainer.cnncopy \
42
      --package-path ./trainer \
43
      --region $REGION \
44
45
      --train-file gs://$BUCKET NAME/data/random shapes.pkl
```

11.6.17:

There is an memory error when running on Google Cloud's regular CPU after one set of 10 epochs for the first half of the dataset. There is not enough memory allocated and training took 1 hour, which is too slow. I decided to use yaml configuration to run on a single NVIDIA K80 GPU processor on Google Cloud Compute Engine.

11.7.17:

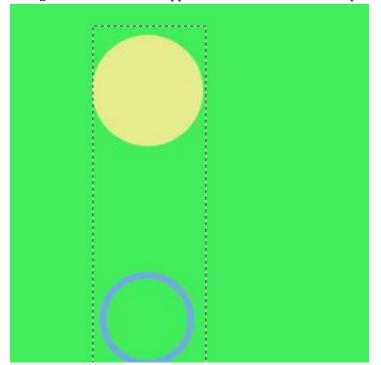
I executed this with no errors in Google Cloud with GPU computing on a validation set 1000 images and training set of 6000 images with roughly 60 percent accuracy, 3 percent error rate in 3 series of 10 epochs per training set each. The learning model is able to be saved. Google Cloud automatically plots the gradient on Tensorboard. The reason the error rate is so high and accuracy is low is because there are alot of background samples that the CNN intakes as pool sizes. Background colored samples are data that contains no linear information - unimportant numpy array figures. so when the network does the maxpool of background samples near the 'important line samples', if the background samples are in greater distributation than the amount of important line samples, maxpool will label that area as background sample which makes the neurons increase the weights for backgrounds instead of the contour images itself.

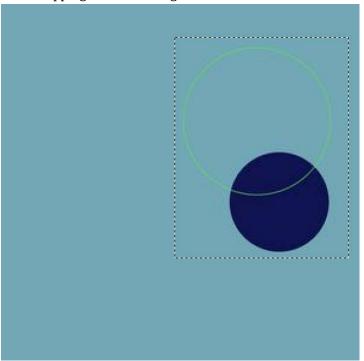
11.8.17:

I increased the y-label output from 3 classes to 4 classes. Keras does the automatic shuffle at every epoch in fit_generator. I changed the architecture of the CNN, add drop out layers that might drop out neurons that have no data of contour characteristics being drawn or do some cropping of batches that do not consist of contour information beforehand. I increased the pool size of the CNN and changed it from adam optimizer to rms optimizer. The CNN will do fit the generator model from data augmentation in 20 epochs with validation and training inputs inputted. I also implemented the validation set correctly during the fitting of the network with real data augmentation. The CNN does poorly during training, with an accuracy of 59 percent and 6 percent loss. This is because I used 3,000 images to train the dataset, which is 1/3 of the total training set, which might not contain evenly distributed images of each type of shape. I reduced the total

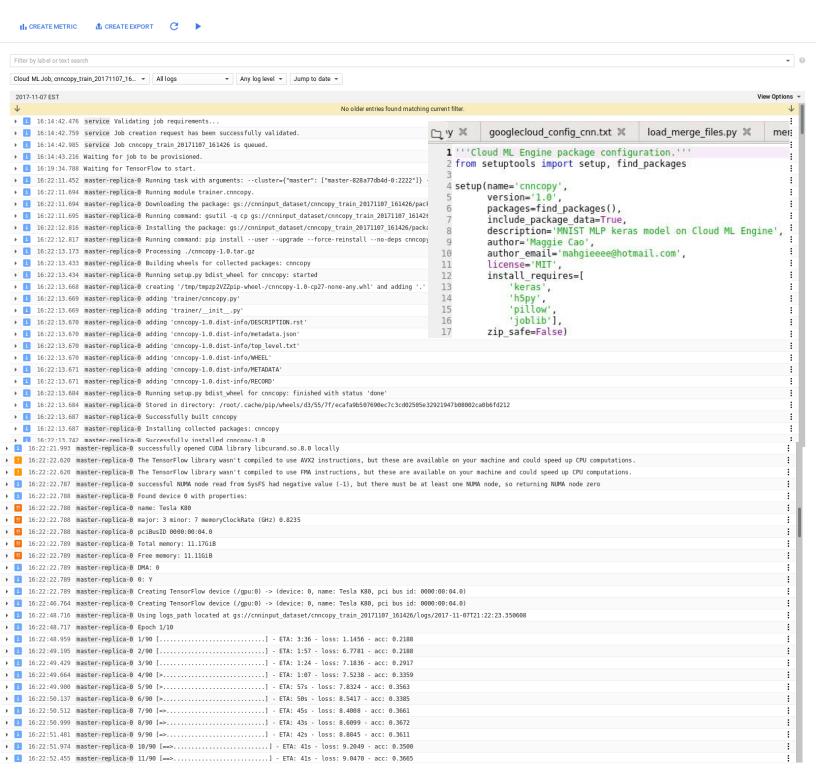
training set by a third because I want to focus on getting the architecture of the CNN right and there is memory error at the Tesla K80 GPU from the loading of the images since the validation data increased by twice as much as the previous one. **11.9.17**:

Trying to figure out how to redesign the architecture of my CNN by looking back on the research I did in Neural Network Design. I also need to create my own data generator (augmentation) function that crops large scaled images to reduce unnecessary background sampling of images in Pooling. I don't want to separate the contours and filling of the images from the background because the background plays an important part in the composition of the entire image object. Such images that need to be cropped, where the dotted lines represent the cropping location, in a generator function are:





When I ran the same code in google cloud, the machine only computed the 32 epochs for the first half of the dataset in the google cloud's regular machines, the machine returns an exit status of -9, which means that the program exited because there isn't enough memory in the machine to execute the program. I need to use a single GPU to run a simple convolutional neural network. I had to request for a quota of 2 Tesla K80 GPUs in the east-1 area from google cloud. The configuration file and output from google cloud to run using nvidia's GPU in the cloud is below. The above images represent a successful output running using GPU processing. The standard-gpu configuration only uses 1 GPU, the complex model uses 4 GPUs. There is no custom model for 2 GPUs. I will need to use google cloud's VM instance to create a machine running in 2 GPUS, install the necessary keras, tensorflow, nvidia dependencies on the machine if I decide to use 2 GPUS. But, for now, it is important to focus on getting a working architecture for the network. So, I'll use 1 GPU unless absolutely necessary for the network to develop a good generalization model. use a VM instance in the cloud. Below describes the commands to run the python code in google cloud. It requires an empty __init__.py file, setup.py file and remote linux commands.

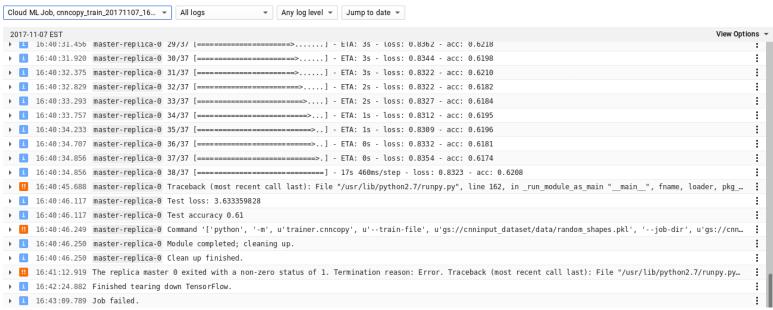


This network is compiled using CUDA's K80 GPU. I learned that the output of the losses are wrong for the first successful compilation of the convolutional neural network in the google cloud. According to http://cs231n.github.io/, the starting loss of the CNN is -ln(1/num_of_classes) . For 3 classes, -ln(0.33) =1.1086626245, which should be the starting loss of the network. However, my starting loss ranges from 6-8%. The output of the program is below, the

The output of the program is below, the compilation error is because of the syntax error in model.save so I wasn't able to save the model in the cloud directory. The test loss (3.63) is fewer than the next keras output as well as a higher accuracy (61%) than the next output, because the next version of the CNN file is being trained on 3033 images in total (1/3 of the total training set due to memory error since I increased the number of output classes from 3 to 4). The validation data

•	i	16:22:52.946	master-replica-0	12/90	[==>] - ETA: 40s - loss: 9.0906 - acc: 0.369	98
٠	i	16:22:53.441	master-replica-0	13/90	[===>] - ETA: 39s - loss: 9.1079 - acc: 0.37	26
•	i	16:22:53.934	master-replica-0	14/90	[===>] - ETA: 39s - loss: 8.8979 - acc: 0.386	84
•	i	16:22:54.423	master-replica-0	15/90	[===>] - ETA: 38s - loss: 8.7307 - acc: 0.400	90
•	i	16:22:54.905	master-replica-0	16/90	[====>] - ETA: 37s - loss: 8.5082 - acc: 0.416	50
•	i	16:22:55.391	master-replica-0	17/90	[===>] - ETA: 37s - loss: 8.5411 - acc: 0.41	36
•	i	16:22:55.887	master-replica-0	18/90	[====>] - ETA: 36s - loss: 8.3572 - acc: 0.414	49
Þ	i	16:22:56.387	master-replica-0	19/90	[====>] - ETA: 36s - loss: 8.0705 - acc: 0.419	94
•	i	16:22:56.874	master-replica-0	20/90	[====>] - ETA: 35s - loss: 7.7490 - acc: 0.414	41
Þ	i	16:22:57.371	master-replica-0	21/90	[====>] - ETA: 35s - loss: 7.4609 - acc: 0.40	77
•	i	16:22:57.868	master-replica-0	22/90	[=====>] - ETA: 34s - loss: 7.1898 - acc: 0.40	34
Þ	i	16:22:58.349	master-replica-0	23/90	[====>] - ETA: 34s - loss: 6.9293 - acc: 0.400	22
•	i	16:22:58.824	master-replica-0	24/90	[====>] - ETA: 33s - loss: 6.6899 - acc: 0.399	97
Þ	i	16:22:59.318	master-replica-0	25/90	[=====>] - ETA: 32s - loss: 6.4653 - acc: 0.40	12
•	i	16:22:59.802	master-replica-0	26/90	[=====>] - ETA: 32s - loss: 6.2582 - acc: 0.400	26
•	i	16:23:00.281	master-replica-0	27/90	[=====>] - ETA: 31s - loss: 6.0662 - acc: 0.400	28
•	i	16:23:00.775	master-replica-0	28/90	[=====>] - ETA: 31s - loss: 5.8884 - acc: 0.40	51
•	i	16:23:01.258	master-replica-0	29/90	[=====>] - ETA: 30s - loss: 5.7216 - acc: 0.40	52
•	i	16:23:01.814	master-replica-0	30/90	[=====>] - ETA: 30s - loss: 5.5683 - acc: 0.400	21
•	i	16:23:02.054	master-replica-0	31/90	[======>] - ETA: 29s - loss: 5.4356 - acc: 0.400	20
•	i	16:23:02.537	master-replica-0	32/90	[======>] - ETA: 28s - loss: 5.3016 - acc: 0.40	31
•	i	16:23:03.021	master-replica-0	33/90	[======>] - ETA: 28s - loss: 5.1739 - acc: 0.400	23
•	i	16:23:03.514	master-replica-0	34/90	[======>:] - ETA: 27s - loss: 5.0556 - acc: 0.390	87
•	i	16:23:03.997	master-replica-0	35/90	[=======>] - ETA: 27s - loss: 4.9434 - acc: 0.394	45
•	i	16:23:04.482	master-replica-0	36/90	[=====>] - ETA: 26s - loss: 4.8363 - acc: 0.39	74
•	i	16:23:04.966	master-replica-0	37/90	[=====>] - ETA: 26s - loss: 4.7379 - acc: 0.39	59
•	i	16:23:05.447	master-replica-0	38/90	[=====>:] - ETA: 25s - loss: 4.6428 - acc: 0.392	29
٠	i	16:23:05.931	master-replica-0	39/90	[======>] - ETA: 25s - loss: 4.5538 - acc: 0.392	25
•	i	16:23:06.423	master-replica-0	40/90	[=======>] - ETA: 24s - loss: 4.4677 - acc: 0.389	97

was organized wrongly in this network because it was just fit into the network. Training should consist of a validation set and training set. This version of the CNN is trained using stochastic gradient training on a total of 6200 images, with half



the dataset in one generator and the other half in another generator, fitting into the CNN using real-data augmentation. The validation data, consisting of roughly 1800 images, isn't incorporated with the 2 training generators. The validation data is just incorporated separately as an additional generator. This just means that Keras will treat it as an additional input batch of data. Below is my CNN file in Keras of the above output:

```
8 def train_model(loader = 'random_shapes.pkl', job_dir = './', **args):
           train_model(loader = 'random_shapes.pkl', job_dir = './', **args):
    from keras.models import Sequential
    from keras.layers import Conv2D
    from keras.layers import MaxPooling2D
    from keras.layers import Platten
    from keras.layers import Dense
    from keras.preprocessing.image import ImageDataGenerator
    from datetime import datetime # for filename conventions
    from tensorflow.python.lib.io import file_io # for better file I/0
    #import hopy # for saving the model
                                                                                                                                                97
                                                                                                                                                99
                                                                                                                                               100
                                                                                                                                               101
                                                                                                                                               102
                                                                                                                                               103
                                                                                                                                               104
105
106
           #import h5py # for saving the model
import joblib
18
                                                                                                                                               107
19
                                                                                                                                               108
                                                                                                                                                            batches2 = 0
           #set the loggining path for ML Engine logging to storage bucket
logs_path = job_dir + '/logs/' + datetime.now().isoformat()
print('Using logs_path located at {}'.format(logs_path))
                                                                                                                                               109
21
23
            with open(loader, 'rb') as f:
    save = joblib.load(f)
                                                                                                                                               114
                   train_shape_dataset = save['train_shape_dataset']
                  train_shape_dataset = save['train_shape_dataset']
train_y_dataset = save['train_y_dataset']
train_shape_halfdataset = save['train_y_halfdataset']
train_y_halfdataset = save['train_y_halfdataset']
#test_shape_dataset = save['test_shape_dataset']
#test_y_dataset = save['test_y_dataset']
#rectangle_ylabel = save['rectangle_ylabel']
del save  # hint to help gc free up memory
                                                                                                                                               116
28
                                                                                                                                               117
                                                                                                                                               118
30
                                                                                                                                               119
                                                                                                                                               120
121
32
                                                                                                                                               123
             # Initialising the CNN, adding a layer
                                                                                                                                               124
            classifier = Sequential()
37
                                                                                                                                               126
38
            # Step 1 - Convolution
            classifier.add(Conv2D(32, (3, 3), input_shape = (300, 300, 3), activation 27
39
40
             # Step 2 - Pooling
                                                                                                                                               129
41
                                                                                                                                               130
            classifier.add(MaxPooling2D(pool size = (2, 2)))
                                                                                                                                               131
             # Adding a second convolutional layer
            classifier.add(Conv2D(32, (3, 3), activation = 'relu'))
            classifier.add(MaxPooling2D(pool_size = (2, 2)))
48
           # Step 3 - Flattening
classifier.add(Flatten())
49
          #Dense function is used to add a fully connected layer at the end # Step 4 - Full connection
         classifier.add(Dense(units = 128, activation = 'relu'))
classifier.add(Dense(units = 3, activation = 'softmax'))
         # Compiling the CNN #change 'binary_crossentropy to categorical'
classifier.compile(optimizer = 'adam', loss = 'categorical_crossentropy', metrics = ['accuracy'])
          # Part 2 - Fitting the CNN to the images
         #augmentation configuration for rescaling test images
         test_datagen = ImageDataGenerator(rescale = 1./255)
         #circle_train = np.array(circle_train)
#circle_ylabel = np.array(circle_ylabel)
         datagen.fit(train_shape_dataset)
         crain_y_dataset,
batch_size = 32,
save_to_dir = "shapes/train/",
save_prefix = "shapes",
save_format = "jpeg"):
#datagen.fit(x_batch) #or (x_batch, y_batch for output of classes)
classifier.fit(x_batch, y_batch)
batches += 1
                 if batches >= len(train_shape_dataset) / 32:
    break #without break, generator will loop indefinitely
         datagen.fit(train_shape_halfdataset)
         batches1 = 0
         #flow() creates batches of randomly transformed images
for x_batch1, y_batch1 in datagen.flow(train_shape_halfdataset,
                                                                            train_y_halfdataset,
```

I don't know how the memory error occurred during pickle loading in the next version of the CNN. The data input that failed on memory error consists of 2100 circles, 2450 squares, 2100 triangles and 2450 squares for the training set, which is a total of 9100 images. The validation set contains 900 circles, 1050 rectangles, 1050 squares and 900 triangles, which is a total of 3900 images. The loading of the validation set is probably where the memory error occurred because joblib won't be able to load a large numpy array of images greater than approximately 3500 images. In addition, the loss is greater and accuracy is less because this versions uses rmsprop, where the gradient is computed on a batch of data instead of for every input data. Rmsprop optimizer divides the gradient by a running average of its recent magnitude. According to fchollet, the designer of Keras, it is recommended to leave the parameters of the optimizer at their default values, except the learning rate which could be freely tuned. Fchollet also says that this optimizer is usually a good choice for recurrent neural networks, but I'm using a CNN network so I will go back to either Adam or SGD as optimizer. The test dataset consists of 2000 images for each shape. Here is the modification of the first Keras code:

```
18 def train_model(train_file = 'gs://cnninput_dataset/data/random_shapes.pkl',
19 job_dir = './',
                                                                                                                                                          # is this enough to test the network correctly? if you want a more manual # representation of fitting the input data use for loop
                                **args):
          # set the loggining path for ML Engine logging to storage bucket
logs_path = job_dir + '/logs/' + datetime.now().isoformat()
print('Using logs_path located at {}'.format(logs_path))
                                                                                                                                                          validate_datagen.fit(validate_shape_dataset)
                                                                                                                                                          validate_generator = datagen.flow(validate_shape_dataset,
                                                                                                                                             106
                                                                                                                                                                                                                  validate_y_dataset,
batch_size = 32)
           # need tensorflow to open file descriptor in order for google cloud to
                                                                                                                                             109
                                                                                                                                                         # the code below fits the training data that is loaded by pickle file
# to prevent memory error, 1/2 of the number of data inputs are feed first
# an epoch define the input being run once from
# the architecture of the cnn is:
          # process it (instead of 'with open(loader, 'rb' as f:')
with file_io.FileIO(train_file, mode='r') as f:
                                                                                                                                             112
                  # joblib loads compressed files consistenting of large datasets
                 # efficiently.
save = joblib.load(f)
                                                                                                                                             114
115
                 save = jobilib.toad(f)
train_shape_dataset = save['train_shape_dataset']
train_y_dataset = save['train_y_dataset']
#train_shape_halfdataset = save['train_shape_halfdataset']
#train_y_halfdataset = save['train_y_halfdataset']
validate_shape_dataset = save['validate_shape_dataset']
                                                                                                                                             117
118
                                                                                                                                                          # -> 3 neurons for output layer -> softmax
                                                                                                                                             120
                  validate y_dataset = save['validate_y_dataset']
del save # hint to help gc free up memory
                                                                                                                                             123
124
          # Initialising the CNN by adding a simple sequential layer
classifier = Sequential()
                                                                                                                                             125
126
127
           # Sequential layer consists of Convolution of type 3 by 3 convolutional
# window with 32 output filters(dimensionality of output space) for each
           # input image uses reLU layers, which is a
          # ''a nonlinear layer, network with relu is trained faster without creating # a decrease in accuracy @ adeshpande3.github.io'' classifier.add(Conv2D(32, (3, 3), input_shape = (300, 300, 3), activation =
                                                                                                                                             134
135
          # Step 2:
# Max Pooling downsamples the number pixels per neuron and create a max
           # number that describes those features in a pool_size of 2 by 2 # change pool size from (2,2) to (8,8) to (4,4)
                                                                                                                                             137
                                                                                                                                             138
           classifier.add(MaxPooling2D(pool size = (4, 4)))
                                                                                                                                             140
141
          # Adding a second convolutional layer, which is the same as the first one
classifier.add(Conv2D(32, (3, 3), activation = 'relu'))
# change pool size from (2,2) to (8,8)
                                                                                                                                              143
          classifier.add(MaxPooling2D(pool size = (4, 4)))
# Dropout layers at the second convolutional layer before flattening
classifier.add(Dropout(0.25))
                                                                                                                                              144
                                                                                                                                              145
                                                                                                                                              146
                                                                                                                                                            #evaluate the model
          # Step 3: Flattening the convolutional layers for input into a fully # connected layer
                                                                                                                                              148
                                                                                                                                             149
150
          classifier.add(Flatten())
                                                                                                                                                                                                             verbose = 0)
                                                                                                                                                          print ("Test loss: ", score[0])
print ("Test accuracy", score[1])
          # Fully connected: Dense function is used to add a fully connected
# 3 layer perceptron at the end
classifier.add(Dense(units = 128, activation = 'relu'))
                                                                                                                                              153
154
155
                                                                                                                                                           classifier.save('model.h5')
          # dropout at the first layer perceptron
classifier.add(Dropout(0.25))
                                                                                                                                              156
          # adding second hidden layer - remove if accuracy decreases or loss increases
classifier.add(Dense(units = 128, activation = 'relu'))
                                                                                                                                              158
                                                                                                                                                                        output_f.write(input_f.read())
           classifier.add(Dropout(0.35))
          # softmax classifier as an activation from the last layer perceptron
# units represent number of output classes
# the output classes are triangle, rectangle, square, circle
classifier.add(Dense(units = 4, activation = 'softmax'))
                                                                                                                                              160
                                                                                                                                              163
          # Compiling the CNN:
          # check if optimizer adam is good, categorical crossentropy is for
# multi-class network, multilabel with intersection needs binary_crossentropy
                                                                                                                                              165
                                                                                                                                                           parser.add_argument('-
                                                                                                                                                                                                  -job-dir
          # and sigmoid activations
# change from adam to rmsprop
                                                                                                                                              167
                                                                                                                                             168
                                                                                                                                                           args = parser.parse_args()
          arguments = args.__dict
                                                                                                                                                           train model(**arguments)
          # Part 2:
          # Feeding CNN the input images and fitting the CNN
          # CNN uses data augmentation configuration to prevent overfitting
# datagen augmentation is for training data input
          datagen = ImageDataGenerator(rescale = 1./255,
                                                          shear_range = 0.2,
zoom_range = 0.2,
horizontal_flip = True)
                                                                                                                                           indicated in the output of the CNN.
```

augmentation configuration for rescaling images used for validation

validate datagen = ImageDataGenerator(rescale = 1./255)

```
# 2DConv -> ReLU -> MaxPool -> 2DConv -> ReLU -> MaxPool -> Flatten() ->
# Fully connected 2-layer neural network
# 128 neurons for the first layer -> ReLU -> 128 for hidden layer -> ReLU
   compute quantities required for featurewise normalization
#datagen.fit(train_shape_dataset)
#early_stopping = EarlyStopping(monitor = 'val_loss', patience = 2)
# fits the model on batches with real-time data augmentation
train_generator = datagen.flow(train_shape_dataset,
validation_data = validate_generator,
                                       validation_steps = 300)
'''#early stopping prevent overfitting after the second half
early_stopping = EarlyStopping(monitor = 'val_loss', patience = 2)
# feed the same data generator the other half of the dataset
datagen.fit(train_shape_halfdataset)
train_generator_half = datagen.flow(train_shape_halfdataset,
                                                        train_y_halfdataset,
                                                        batch_size = 32)
classifier.fit_generator(train_generator_half,
steps_per_epoch = len(train_shape_halfdataset) / 32,
epochs = 20,
callbacks = [early_stopping],
                                          validation_data = validate_generator,
validation_steps = 300)'''
 score = classifier.evaluate(validate shape dataset,
                                               validate_y_dataset,
 # Save the model to the Cloud Storage bucket's jobs directory
with file_io.FileIO('model.h5', mode='r') as input_f:
    with file_io.FileIO(job_dir + '/model.h5', mode='w+') as output_f:
 _name_ == '__main__':
# Parse the input arguments for common Cloud ML Engine options
 help='Cloud storage bucket to export the model')
```

The loss at the beginning of training for this CNN modification is correct because $-\ln(1/4) = 1.3862943611$, which is very close as

```
34/94 [======>:....] - ETA: 29s - loss: 1.3910 - acc: 0.2651
Here is the output from Google Cloud Machine Learning Engine executed on November 9<sup>th</sup>, 2017.
E Using TensorFlow backend.
                                                                          35/94 [=======>......] - ETA: 29s - loss: 1.3909 - acc: 0.2646
                                                                          36/94 [=======>.....] - ETA: 28s - loss: 1.3908 - acc: 0.2660
I successfully opened CUDA library libcublas.so.8.0 locally
I successfully opened CUDA library libcudnn.so.5 locally
                                                                          37/94 [=======>......] - ETA: 28s - loss: 1.3905 - acc: 0.2672
                                                                          38/94 [=======>:....] - ETA: 27s - loss: 1.3901 - acc: 0.2676
I successfully opened CUDA library libcufft.so.8.0 locally
I successfully opened CUDA library libcuda.so.1 locally
                                                                          39/94 [=======>:.....] - ETA: 27s - loss: 1.3900 - acc: 0.2695
I successfully opened CUDA library libcurand.so.8.0 locally
                                                                          40/94 [=======>.....] - ETA: 26s - loss: 1.3897 - acc: 0.2698
                                                                          41/94 [=======>.....] - ETA: 26s - loss: 1.3892 - acc: 0.2701
W The TensorFlow library wasn't compiled to use AVX2 instructions, but these are available on your machine and
                                                                          42/94 [========>.....] - ETA: 25s - loss: 1.3885 - acc: 0.2711
could speed up CPU computations.
W The TensorFlow library wasn't compiled to use FMA instructions, but these are available on your machine and
                                                                          43/94 [========>:....] - ETA: 25s - loss: 1.3887 - acc: 0.2728
                                                                          44/94 [========>.....] - ETA: 24s - loss: 1.3878 - acc: 0.2751
could speed up CPU computations.
                                                                          45/94 [========>.....] - ETA: 24s - loss: 1.3864 - acc: 0.2767
I successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so
                                                                          46/94 [========>:....] - ETA: 23s - loss: 1.3869 - acc: 0.2747
returning NUMA node zero
                                                                          47/94 [========>.....] - ETA: 23s - loss: 1.3866 - acc: 0.2775
I Found device 0 with properties:
E name: Tesla K80
                                                                          48/94 [=========>.....] - ETA: 22s - loss: 1.3867 - acc: 0.2756
E major: 3 minor: 7 memoryClockRate (GHz) 0.8235
                                                                          49/94 [========>.....] - ETA: 22s - loss: 1.3862 - acc: 0.2758
                                                                          50/94 [========>.....] - ETA: 21s - loss: 1.3869 - acc: 0.2740
E pciBusID 0000:00:04.0
                                                                          51/94 [=========>.....] - ETA: 21s - loss: 1.3871 - acc: 0.2735
E Total memory: 11.17GiB
                                                                          E Free memory: 11.11GiB
                                                                          53/94 [=======>:::...] - ETA: 20s - loss: 1.3867 - acc: 0.2779
I DMA: 0
                                                                          54/94 [==========>.....] - ETA: 19s - loss: 1.3867 - acc: 0.2780
I 0: Y
I Creating TensorFlow device (/gpu:0) -> (device: 0, name: Tesla K80, pci bus id: 0000:00:04.0)
                                                                          55/94 [=========>......] - ETA: 19s - loss: 1.3862 - acc: 0.2798
                                                                          56/94 [==========>.....] - ETA: 18s - loss: 1.3862 - acc: 0.2792
I Creating TensorFlow device (/gpu:0) -> (device: 0, name: Tesla K80, pci bus id: 0000:00:04.0)
I Using logs_path located at gs://cnninput_dataset/cnncopy_train_20171109_022628/logs/2017-11-
                                                                          57/94 [=========>:.....] - ETA: 18s - loss: 1.3858 - acc: 0.2804
09T07:26:51.959549
                                                                          59/94 [==========>...........] - ETA: 17s - loss: 1.3853 - acc: 0.2804
Epoch 1/20
                                                                          60/94 [============>......] - ETA: 16s - loss: 1.3848 - acc: 0.2825
1/94 [......] - ETA: 2:51 - loss: 1.4025 - acc: 0.2188
                                                                          61/94 [============>..........] - ETA: 16s - loss: 1.3847 - acc: 0.2835
2/94 [......] - ETA: 1:31 - loss: 1.3896 - acc: 0.2500
                                                                          62/94 [===========>......] - ETA: 15s - loss: 1.3838 - acc: 0.2845
3/94 [......] - ETA: 1:04 - loss: 1.3566 - acc: 0.2917
                                                                          63/94 [===========>..........] - ETA: 15s - loss: 1.3838 - acc: 0.2849
4/94 [>.....] - ETA: 51s - loss: 1.4008 - acc: 0.2734
                                                                          5/94 [>......] - ETA: 46s - loss: 1.4005 - acc: 0.2750
6/94 [>.....] - ETA: 45s - loss: 1.4118 - acc: 0.2500
                                                                          66/94 [=============>.......] - ETA: 13s - loss: 1.3832 - acc: 0.2843
7/94 [=>.....] - ETA: 44s - loss: 1.4092 - acc: 0.2500
8/94 [=>.....] - ETA: 43s - loss: 1.4071 - acc: 0.2500
                                                                          9/94 [=>.....] - ETA: 43s - loss: 1.4063 - acc: 0.2604
                                                                          10/94 [==>.....] - ETA: 42s - loss: 1.4048 - acc: 0.2531
11/94 [==>.....] - ETA: 41s - loss: 1.3998 - acc: 0.2614
                                                                          12/94 [==>.....] - ETA: 41s - loss: 1.4031 - acc: 0.2578
13/94 [===>.....] - ETA: 40s - loss: 1.4018 - acc: 0.2644
                                                                          14/94 [===>.....] - ETA: 39s - loss: 1.3994 - acc: 0.2723
                                                                          15/94 [===>.....] - ETA: 39s - loss: 1.3985 - acc: 0.2729
16/94 [====>.....] - ETA: 38s - loss: 1.3972 - acc: 0.2773
                                                                          17/94 [====>.....] - ETA: 38s - loss: 1.3968 - acc: 0.2757
                                                                          18/94 [====>.....] - ETA: 37s - loss: 1.3973 - acc: 0.2691
                                                                          19/94 [====>.....] - ETA: 37s - loss: 1.3966 - acc: 0.2671
20/94 [====>.....] - ETA: 36s - loss: 1.3944 - acc: 0.2694
                                                                          21/94 [====>.....] - ETA: 35s - loss: 1.3923 - acc: 0.2714
                                                                          22/94 [====>.....] - ETA: 35s - loss: 1.3947 - acc: 0.2676
                                                                          23/94 [=====>.....] - ETA: 34s - loss: 1.3939 - acc: 0.2709
24/94 [=====>.....] - ETA: 34s - loss: 1.3934 - acc: 0.2726
                                                                          25/94 [=====>.....] - ETA: 33s - loss: 1.3936 - acc: 0.2680
                                                                          26/94 [======>.....] - ETA: 33s - loss: 1.3926 - acc: 0.2709
                                                                          27/94 [======>.....] - ETA: 32s - loss: 1.3919 - acc: 0.2690
28/94 [======>.....] - ETA: 32s - loss: 1.3923 - acc: 0.2683
                                                                          29/94 [======>.....] - ETA: 31s - loss: 1.3920 - acc: 0.2655
                                                                          30/94 [======>.....] - ETA: 31s - loss: 1.3912 - acc: 0.2650
31/94 [======>.....] - ETA: 30s - loss: 1.3910 - acc: 0.2665
                                                                          32/94 [======>.....] - ETA: 30s - loss: 1.3908 - acc: 0.2680
33/94 [======>.....] - ETA: 29s - loss: 1.3905 - acc: 0.2674
```

-	======================================		======>] - ETA: 15s - loss: 1.3159 - acc: 0.3583
-	=========>.] - ETA: 0s - loss: 1.3710 - acc: 0.2983		=======>] - ETA: 15s - loss: 1.3159 - acc: 0.3575
-	======================================		=======>] - ETA: 14s - loss: 1.3164 - acc: 0.3567
.3217			=======>] - ETA: 14s - loss: 1.3159 - acc: 0.3575
ooch 2/20			=======>] - ETA: 14s - loss: 1.3158 - acc: 0.3578
=			=======>>] - ETA: 13s - loss: 1.3152 - acc: 0.3576
-			=======>] - ETA: 13s - loss: 1.3160 - acc: 0.3564
-			=======>>] - ETA: 13s - loss: 1.3142 - acc: 0.3562
-] - ETA: 13s - loss: 1.2907 - acc: 0.4141	-	=======>] - ETA: 12s - loss: 1.3132 - acc: 0.3569
-] - ETA: 13s - loss: 1.3040 - acc: 0.4062	-	=======>] - ETA: 12s - loss: 1.3112 - acc: 0.3577
-		-	=======>] - ETA: 11s - loss: 1.3109 - acc: 0.3556
] - ETA: 12s - loss: 1.3309 - acc: 0.3750		========>] - ETA: 11s - loss: 1.3127 - acc: 0.3540
-] - ETA: 12s - loss: 1.3210 - acc: 0.3867	-	========>] - ETA: 11s - loss: 1.3129 - acc: 0.3543
-		-	========>] - ETA: 10s - loss: 1.3124 - acc: 0.3528
-		-	========>] - ETA: 10s - loss: 1.3116 - acc: 0.3540
-		-	========>] - ETA: 9s - loss: 1.3121 - acc: 0.3526
-		=	========>] - ETA: 9s - loss: 1.3127 - acc: 0.3494
-			========>] - ETA: 9s - loss: 1.3123 - acc: 0.3510
		-	
-		=	
-		-	
-		=	
-] - ETA: 16s - loss: 1.3164 - acc: 0.3569] - ETA: 17s - loss: 1.3168 - acc: 0.3563	-	
-	•	-	======================================
-] - ETA: 18s - loss: 1.3185 - acc: 0.3542		·
-] - ETA: 18s - loss: 1.3201 - acc: 0.3537] - ETA: 19s - loss: 1.3239 - acc: 0.3519	=	======================================
-	•	=	
-] - ETA: 19s - loss: 1.3231 - acc: 0.3529] - ETA: 19s - loss: 1.3233 - acc: 0.3550	-	======================================
-] - ETA: 195 - 1055: 1.3233 - acc: 0.3550	-	======================================
-	:] - ETA: 20s - Ioss: 1.3211 - acc. 0.3322	-	======================================
-		-	======================================
-	:>] - ETA: 20s - loss: 1.3291 - acc: 0.3416	-	======================================
-	>	-	========:] - ETA: 2s - loss: 1.3068 - acc: 0.3478
-	:>	-	=======:] - ETA: 1s - loss: 1.3080 - acc: 0.3467
-	=>] - ETA: 20s - loss: 1.3233 - acc: 0.3535	-	======================================
-	=>] - ETA: 20s - loss: 1.3218 - acc: 0.3532	-	======================================
-	=>] - ETA: 20s - loss: 1.3186 - acc: 0.3575	-	======================================
-	==>		======================================
	==>] - ETA: 20s - loss: 1.3237 - acc: 0.3498	0.3800	,
-	==>] - ETA: 20s - loss: 1.3223 - acc: 0.3539	Epoch 3/20	
-	:===>] - ETA: 20s - loss: 1.3194 - acc: 0.3569	1/94 [] - ETA: 15s - loss: 1.3007 - acc: 0.4375
-	:===>] - ETA: 20s - loss: 1.3183 - acc: 0.3542	-] - ETA: 15s - loss: 1.2292 - acc: 0.4375
-	===>] - ETA: 19s - loss: 1.3132 - acc: 0.3586		
	===>] - ETA: 14s - loss: 1.2406 - acc: 0.3906
2/94 [======	====>] - ETA: 19s - loss: 1.3201 - acc: 0.3579	5/94 [>] - ETA: 13s - loss: 1.2271 - acc: 0.3812
3/94 [======	====>] - ETA: 19s - loss: 1.3181 - acc: 0.3612	6/94 [>] - ETA: 13s - loss: 1.2247 - acc: 0.3802
	====>] - ETA: 19s - loss: 1.3178 - acc: 0.3629] - ETA: 13s - loss: 1.2183 - acc: 0.3839
	=====>] - ETA: 18s - loss: 1.3157 - acc: 0.3646	8/94 [=>] - ETA: 12s - loss: 1.2190 - acc: 0.3867
	=====>] - ETA: 18s - loss: 1.3162 - acc: 0.3628	9/94 [=>] - ETA: 12s - loss: 1.2302 - acc: 0.3958
	=====>] - ETA: 18s - loss: 1.3156 - acc: 0.3617	10/94 [==>] - ETA: 12s - loss: 1.2274 - acc: 0.3937
	=====>] - ETA: 17s - loss: 1.3155 - acc: 0.3607	11/94 [==>	
-	=====>] - ETA: 17s - loss: 1.3161 - acc: 0.3610		
-	======>] - ETA: 17s - loss: 1.3130 - acc: 0.3627	-] - ETA: 11s - loss: 1.2288 - acc: 0.3942
	:=====>] - ETA: 16s - loss: 1.3134 - acc: 0.3624	=] - ETA: 11s - loss: 1.2555 - acc: 0.3839
	:=====>] - ETA: 16s - loss: 1.3128 - acc: 0.3620	=] - ETA: 12s - loss: 1.2673 - acc: 0.3792
	======>] - ETA: 16s - loss: 1.3173 - acc: 0.3605	=] - ETA: 13s - loss: 1.2665 - acc: 0.3789
	======>	=	

```
18/94 [====>.....] - ETA: 16s - loss: 1.2606 - acc: 0.3837
19/94 [====>.....] - ETA: 17s - loss: 1.2560 - acc: 0.3849
20/94 [====>.....] - ETA: 17s - loss: 1.2544 - acc: 0.3844
21/94 [====>.....] - ETA: 18s - loss: 1.2544 - acc: 0.3839
22/94 [====>.....] - ETA: 18s - loss: 1.2486 - acc: 0.3864
23/94 [=====>.....] - ETA: 19s - loss: 1.2472 - acc: 0.3859
24/94 [=====>.....] - ETA: 19s - loss: 1.2480 - acc: 0.3854
25/94 [=====>.....] - ETA: 20s - loss: 1.2542 - acc: 0.3787
26/94 [=====>.....] - ETA: 20s - loss: 1.2524 - acc: 0.3798
27/94 [======>.....] - ETA: 20s - loss: 1.2482 - acc: 0.3785
28/94 [======>.....] - ETA: 20s - loss: 1.2450 - acc: 0.3750
29/94 [======>.....] - ETA: 20s - loss: 1.2427 - acc: 0.3772
30/94 [======>.....] - ETA: 20s - loss: 1.2419 - acc: 0.3833
31/94 [======>.....] - ETA: 20s - loss: 1.2435 - acc: 0.3790
32/94 [======>.....] - ETA: 20s - loss: 1.2415 - acc: 0.3789
33/94 [======>.....] - ETA: 20s - loss: 1.2396 - acc: 0.3788
34/94 [=======>.....] - ETA: 20s - loss: 1.2383 - acc: 0.3778
35/94 [=======>.....] - ETA: 20s - loss: 1.2356 - acc: 0.3768
36/94 [=======>.....] - ETA: 20s - loss: 1.2415 - acc: 0.3741
37/94 [=======>......] - ETA: 20s - loss: 1.2405 - acc: 0.3725
38/94 [=======>.....] - ETA: 20s - loss: 1.2384 - acc: 0.3750
39/94 [=======>.....] - ETA: 20s - loss: 1.2377 - acc: 0.3758
40/94 [=======>:....] - ETA: 20s - loss: 1.2381 - acc: 0.3789
41/94 [=======>:....] - ETA: 19s - loss: 1.2392 - acc: 0.3788
42/94 [=======>.....] - ETA: 19s - loss: 1.2407 - acc: 0.3802
43/94 [========>.....] - ETA: 19s - loss: 1.2415 - acc: 0.3808
44/94 [========>.....] - ETA: 19s - loss: 1.2408 - acc: 0.3814
45/94 [========>.....] - ETA: 18s - loss: 1.2400 - acc: 0.3812
46/94 [========>.....] - ETA: 18s - loss: 1.2392 - acc: 0.3784
47/94 [=======>.....] - ETA: 18s - loss: 1.2352 - acc: 0.3836
48/94 [========>.....] - ETA: 18s - loss: 1.2343 - acc: 0.3861
49/94 [=======>:.....] - ETA: 17s - loss: 1.2338 - acc: 0.3865
50/94 [========>.....] - ETA: 17s - loss: 1.2354 - acc: 0.3862
51/94 [=========>......] - ETA: 17s - loss: 1.2324 - acc: 0.3897
52/94 [========>......] - ETA: 16s - loss: 1.2323 - acc: 0.3882
53/94 [========>.....] - ETA: 16s - loss: 1.2330 - acc: 0.3915
55/94 [=========>.....] - ETA: 15s - loss: 1.2344 - acc: 0.3881
56/94 [=========>.....] - ETA: 15s - loss: 1.2340 - acc: 0.3895
57/94 [=========>......] - ETA: 15s - loss: 1.2328 - acc: 0.3893
58/94 [==========>......] - ETA: 14s - loss: 1.2308 - acc: 0.3901
59/94 [========>:.....] - ETA: 14s - loss: 1.2280 - acc: 0.3935
60/94 [=========>.....] - ETA: 14s - loss: 1.2261 - acc: 0.3932
62/94 [=========>:..........] - ETA: 13s - loss: 1.2274 - acc: 0.3942
67/94 [==========>........] - ETA: 11s - loss: 1.2279 - acc: 0.3913
```

```
95/94 [=========================] - 188s 2s/step - loss: 1.2214 - acc: 0.3915 - val_loss: 1.1272 - val_acc:
0.4431
Epoch 4/20
1/94 [.....] - ETA: 16s - loss: 1.0405 - acc: 0.5000
3/94 [.....] - ETA: 14s - loss: 1.0883 - acc: 0.4896
4/94 [>.....] - ETA: 14s - loss: 1.0864 - acc: 0.4609
5/94 [>.....] - ETA: 13s - loss: 1.0686 - acc: 0.4625
6/94 [>.....] - ETA: 13s - loss: 1.0807 - acc: 0.4635
7/94 [=>.....] - ETA: 12s - loss: 1.0831 - acc: 0.4688
8/94 [=>.....] - ETA: 12s - loss: 1.0931 - acc: 0.4531
9/94 [=>.....] - ETA: 12s - loss: 1.1091 - acc: 0.4514
10/94 [==>.....] - ETA: 11s - loss: 1.0966 - acc: 0.4594
11/94 [==>.....] - ETA: 11s - loss: 1.0954 - acc: 0.4517
12/94 [==>.....] - ETA: 11s - loss: 1.0997 - acc: 0.4375
13/94 [===>.....] - ETA: 11s - loss: 1.0926 - acc: 0.4423
14/94 [===>.....] - ETA: 10s - loss: 1.0966 - acc: 0.4327
15/94 [===>.....] - ETA: 11s - loss: 1.0887 - acc: 0.4330
16/94 [====>.....] - ETA: 13s - loss: 1.0804 - acc: 0.4411
17/94 [====>.....] - ETA: 14s - loss: 1.0888 - acc: 0.4409
18/94 [====>.....] - ETA: 15s - loss: 1.0871 - acc: 0.4407
19/94 [====>.....] - ETA: 16s - loss: 1.0821 - acc: 0.4471
20/94 [====>.....] - ETA: 17s - loss: 1.0871 - acc: 0.4435
21/94 [====>.....] - ETA: 17s - loss: 1.0870 - acc: 0.4417
22/94 [====>.....] - ETA: 18s - loss: 1.0884 - acc: 0.4373
23/94 [=====>.....] - ETA: 18s - loss: 1.0873 - acc: 0.4332
24/94 [=====>.....] - ETA: 19s - loss: 1.0986 - acc: 0.4230
25/94 [=====>.....] - ETA: 19s - loss: 1.1035 - acc: 0.4223
26/94 [=====>.....] - ETA: 19s - loss: 1.1002 - acc: 0.4277
27/94 [======>.....] - ETA: 20s - loss: 1.0980 - acc: 0.4292
28/94 [======>.....] - ETA: 20s - loss: 1.0948 - acc: 0.4340
29/94 [======>.....] - ETA: 20s - loss: 1.0880 - acc: 0.4406
30/94 [======>.....] - ETA: 20s - loss: 1.0871 - acc: 0.4384
31/94 [======>.....] - ETA: 20s - loss: 1.0859 - acc: 0.4394
32/94 [=======>......] - ETA: 20s - loss: 1.0823 - acc: 0.4393
33/94 [=======>.....] - ETA: 20s - loss: 1.0814 - acc: 0.4383
34/94 [=======>.....] - ETA: 20s - loss: 1.0887 - acc: 0.4318
35/94 [=======>.....] - ETA: 20s - loss: 1.0990 - acc: 0.4293
36/94 [=======>.....] - ETA: 20s - loss: 1.1002 - acc: 0.4304
37/94 [=======>.....] - ETA: 20s - loss: 1.0999 - acc: 0.4298
38/94 [=======>:.....] - ETA: 19s - loss: 1.1019 - acc: 0.4300
```

9/94 [=======>,	.] - ETA: 19s - loss: 1.0984 - acc: 0.4302	1/94	[] - ETA: 15s - loss: 0.9727 - acc: 0.5625
-		.l - ETA: 19s - loss: 1.0950 - acc: 0.4335		[] - ETA: 15s - loss: 0.9740 - acc: 0.5469
L/94 [======>,,,,,,	.] - ETA: 19s - loss: 1.0976 - acc: 0.4320		[] - ETA: 14s - loss: 1.0323 - acc: 0.5312
_] - ETA: 19s - loss: 1.0961 - acc: 0.4299		[>] - ETA: 14s - loss: 1.0174 - acc: 0.5078
-] - ETA: 19s - loss: 1.0933 - acc: 0.4338		[>] - ETA: 13s - loss: 1.0071 - acc: 0.5062
1/94 [======>] - ETA: 18s - loss: 1.0939 - acc: 0.4338	6/94	[>] - ETA: 13s - loss: 1.0059 - acc: 0.4948
5/94 [======>] - ETA: 18s - loss: 1.0960 - acc: 0.4311		[=>] - ETA: 12s - loss: 1.0033 - acc: 0.4866
/94 [=====>] - ETA: 18s - loss: 1.0967 - acc: 0.4286	8/94	[=>] - ETA: 12s - loss: 0.9894 - acc: 0.4883
/94 [=====>] - ETA: 18s - loss: 1.0972 - acc: 0.4281	9/94	[=>] - ETA: 12s - loss: 0.9813 - acc: 0.5000
/94 [======>] - ETA: 17s - loss: 1.0944 - acc: 0.4296	10/94	[==>] - ETA: 11s - loss: 0.9744 - acc: 0.5031
/94 [=====>] - ETA: 17s - loss: 1.0937 - acc: 0.4310	11/94	[==>] - ETA: 11s - loss: 0.9634 - acc: 0.5142
/94 [======>] - ETA: 17s - loss: 1.0922 - acc: 0.4343	12/94	[==>] - ETA: 11s - loss: 0.9566 - acc: 0.5130
/94 [======>,] - ETA: 16s - loss: 1.0902 - acc: 0.4362	13/94	[===>] - ETA: 11s - loss: 0.9537 - acc: 0.5120
/94 [======>] - ETA: 16s - loss: 1.0879 - acc: 0.4386	14/94	[===>] - ETA: 11s - loss: 0.9671 - acc: 0.5022
/94 [======>,] - ETA: 16s - loss: 1.0859 - acc: 0.4398	15/94	[===>] - ETA: 12s - loss: 0.9876 - acc: 0.4917
/94 [======>,] - ETA: 15s - loss: 1.0852 - acc: 0.4403	16/94	[====>] - ETA: 13s - loss: 1.0053 - acc: 0.4785
/94 [======>,] - ETA: 15s - loss: 1.0839 - acc: 0.4385	17/94	[====>] - ETA: 14s - loss: 1.0006 - acc: 0.4835
/94 [======>,] - ETA: 15s - loss: 1.0849 - acc: 0.4396	18/94	[====>] - ETA: 15s - loss: 0.9983 - acc: 0.4844
/94 [======>,] - ETA: 14s - loss: 1.0983 - acc: 0.4380	19/94	[====>] - ETA: 16s - loss: 0.9977 - acc: 0.4819
′94 [======>] - ETA: 14s - loss: 1.0957 - acc: 0.4396	20/94	[====>] - ETA: 17s - loss: 0.9940 - acc: 0.4813
/94 [======>,] - ETA: 14s - loss: 1.0959 - acc: 0.4401	21/94	[====>] - ETA: 18s - loss: 0.9878 - acc: 0.4851
/94 [======>] - ETA: 13s - loss: 1.0950 - acc: 0.4421	22/94	[====>] - ETA: 18s - loss: 0.9852 - acc: 0.4858
/94 [======>] - ETA: 13s - loss: 1.0949 - acc: 0.4405	23/94	[=====>] - ETA: 18s - loss: 0.9834 - acc: 0.4831
/94 [======>] - ETA: 13s - loss: 1.0918 - acc: 0.4425	24/94	[=====>] - ETA: 19s - loss: 0.9864 - acc: 0.4877
-] - ETA: 12s - loss: 1.0892 - acc: 0.4434		[=====>] - ETA: 19s - loss: 0.9834 - acc: 0.4957
] - ETA: 12s - loss: 1.0893 - acc: 0.4433	26/94	[=====>] - ETA: 19s - loss: 0.9762 - acc: 0.5018
/94 [=======>,] - ETA: 12s - loss: 1.0917 - acc: 0.4408	27/94	[======>] - ETA: 20s - loss: 0.9737 - acc: 0.4995
-] - ETA: 11s - loss: 1.0943 - acc: 0.4412		[=====>] - ETA: 20s - loss: 0.9776 - acc: 0.4928
-		·] - ETA: 11s - loss: 1.0930 - acc: 0.4412		[======>] - ETA: 20s - loss: 0.9749 - acc: 0.4941
-] - ETA: 10s - loss: 1.0901 - acc: 0.4425		[======>] - ETA: 20s - loss: 0.9764 - acc: 0.4933
-		·] - ETA: 10s - loss: 1.0891 - acc: 0.4420		[======>] - ETA: 20s - loss: 0.9733 - acc: 0.4975
-		->] - ETA: 10s - loss: 1.0889 - acc: 0.4414		[=======>] - ETA: 20s - loss: 0.9716 - acc: 0.4995
-		->] - ETA: 9s - loss: 1.0887 - acc: 0.4396		[=======>] - ETA: 20s - loss: 0.9742 - acc: 0.4967
-		->] - ETA: 9s - loss: 1.0869 - acc: 0.4405		[=======>] - ETA: 20s - loss: 0.9761 - acc: 0.4941
		==>] - ETA: 9s - loss: 1.0856 - acc: 0.4404		[=======>] - ETA: 20s - loss: 0.9763 - acc: 0.4916
-		=>] - ETA: 8s - loss: 1.0861 - acc: 0.4387		[=======>] - ETA: 20s - loss: 0.9817 - acc: 0.4883
		==>] - ETA: 8s - loss: 1.0853 - acc: 0.4379		[=======>] - ETA: 20s - loss: 0.9837 - acc: 0.4878
_		==>] - ETA: 7s - loss: 1.0834 - acc: 0.4399		[=====================================
		==>] - ETA: 7s - loss: 1.0822 - acc: 0.4399		[========>] - ETA: 19s - loss: 0.9835 - acc: 0.4892
-		:==>] - ETA: 7s - loss: 1.0821 - acc: 0.4398		[=====================================
		:==>] - ETA: 6s - loss: 1.0802 - acc: 0.4422 :===>] - ETA: 6s - loss: 1.0793 - acc: 0.4433		[=======>] - ETA: 19s - loss: 0.9791 - acc: 0.4936 [=====>] - ETA: 19s - loss: 0.9771 - acc: 0.4959
		===>] - ETA: 5s - loss: 1.0793 - acc. 0.4433		[=====================================
		===>] - ETA: 5s - loss: 1.0770 - acc. 0.4417		[========>] - ETA: 175 - loss: 0.7700 - acc: 0.4700
		:===>] - ETA: 4s - loss: 1.0744 - acc: 0.4442		[========>] - ETA: 16s - loss: 0.9789 - acc: 0.4944
_		====>] - ETA: 4s - loss: 1.0744 - acc: 0.4442		[========>:] - ETA: 103 - loss: 0.7767 - acc: 0.4734
		:===>] - ETA: 4s - loss: 1.0730 - acc: 0.4441		[=======>:] - ETA: 18s - loss: 0.7802 - acc: 0.4722
_		=====>] - ETA: 43 loss: 1.0732 acc: 0.4458		[=====================================
		=====>] - ETA: 3s - loss: 1.0754 - acc: 0.4461		[=====================================
_		=====>] - ETA: 2s - loss: 1.0730 acc: 0.4470		[=====================================
		======>] - ETA: 2s - loss: 1.0736 - acc: 0.4480		[=====================================
		=====>] - ETA: 2s - loss: 1.0730 acc: 0.4489		[=====================================
		=====>] - ETA: 23 - loss: 1.0727 - acc: 0.4505		[=====================================
		======>.] - ETA: 1s - loss: 1.0724 - acc: 0.4524		[=====================================
		======>.] - ETA: 0s - loss: 1.0695 - acc: 0.4522		[=====================================
-		======>.] - ETA: 0s - loss: 1.0665 - acc: 0.4541		[=====================================
-		========] - 188s 2s/step - loss: 1.0646 - acc: 0.4565 - val_loss: 0.9222 - val_acc:		
,,,,, 5043		, 1003 25/300p 1055/100 10 dec. 0.7505 Val_1055/0/222 Val_dec.		[=====================================

60/94 [[=========>] - ETA: 14s - loss: 0.9771 - acc: 0.5024	22/94	[====>] - ETA: 18s - loss: 0.9141 - acc: 0.5312
	[=====================================		[=====>] - ETA: 19s - loss: 0.9107 - acc: 0.5285
	[========>:] - ETA: 13s - loss: 0.9771 - acc: 0.5058		[=====>] - ETA: 19s - loss: 0.9044 - acc: 0.5326
63/94 [[========>:] - ETA: 12s - loss: 0.9768 - acc: 0.5062	25/94	[=====>] - ETA: 20s - loss: 0.9005 - acc: 0.5363
64/94 [[==============>] - ETA: 12s - loss: 0.9751 - acc: 0.5071	26/94	[=====>] - ETA: 20s - loss: 0.8961 - acc: 0.5397
65/94 [[=====================================	27/94	[======>] - ETA: 20s - loss: 0.8902 - acc: 0.5417
66/94 [[=====================================	28/94	[=====>] - ETA: 20s - loss: 0.8860 - acc: 0.5413
	[=====================================	29/94	[======>] - ETA: 20s - loss: 0.8882 - acc: 0.5366
68/94 [[=====================================	30/94	[======>] - ETA: 20s - loss: 0.8894 - acc: 0.5323
	[=====================================	31/94	[======>] - ETA: 20s - loss: 0.8908 - acc: 0.5343
70/94 [[=====================================	32/94	[=======>] - ETA: 20s - loss: 0.8922 - acc: 0.5303
71/94 [[=====================================	33/94	[=======>] - ETA: 20s - loss: 0.8972 - acc: 0.5275
72/94 [[=====================================	34/94	[======>] - ETA: 20s - loss: 0.8968 - acc: 0.5248
73/94 [[=====================================	35/94	[======>:] - ETA: 20s - loss: 0.8930 - acc: 0.5268
74/94 [[=====================================	36/94	[======>:] - ETA: 20s - loss: 0.8909 - acc: 0.5278
75/94 [[=====================================	37/94	[======>:] - ETA: 20s - loss: 0.8891 - acc: 0.5279
76/94 [[=====================================	38/94	[======>:] - ETA: 20s - loss: 0.8883 - acc: 0.5247
77/94 [[=====================================	39/94	[======>:] - ETA: 20s - loss: 0.8917 - acc: 0.5232
78/94 [[=====================================	40/94	[======>:] - ETA: 19s - loss: 0.8917 - acc: 0.5250
79/94 [[=====================================	41/94	[======>:] - ETA: 19s - loss: 0.8936 - acc: 0.5252
80/94 [[=====================================	42/94	[======>:] - ETA: 19s - loss: 0.8918 - acc: 0.5253
81/94 [[=====================================	43/94	[======>:] - ETA: 19s - loss: 0.8885 - acc: 0.5276
82/94 [[=====================================	44/94	[======>:] - ETA: 19s - loss: 0.8846 - acc: 0.5298
	[=====================================	45/94	[=======>:] - ETA: 18s - loss: 0.8868 - acc: 0.5278
84/94 [[=====================================		[=======>:] - ETA: 18s - loss: 0.8976 - acc: 0.5265
85/94 [[=====================================	47/94	[=======>:] - ETA: 18s - loss: 0.8993 - acc: 0.5219
	[=====================================	48/94	[=======>:] - ETA: 18s - loss: 0.9008 - acc: 0.5215
	[=====================================	49/94	[=======>:] - ETA: 17s - loss: 0.8977 - acc: 0.5230
	[=====================================	50/94	[======>:] - ETA: 17s - loss: 0.8961 - acc: 0.5250
_	[=====================================		[======>:] - ETA: 17s - loss: 0.8936 - acc: 0.5263
90/94 [[=====================================	52/94	[========>:] - ETA: 16s - loss: 0.8933 - acc: 0.5270
	[=====================================	53/94	[======>:] - ETA: 16s - loss: 0.8923 - acc: 0.5271
92/94 [[=====================================		[======>:] - ETA: 16s - loss: 0.8911 - acc: 0.5272
93/94 [[=====================================		[=============>] - ETA: 15s - loss: 0.8950 - acc: 0.5267
-	[=====================================		[=============>] - ETA: 15s - loss: 0.8963 - acc: 0.5285
	[==============================] - 187s 2s/step - loss: 0.9621 - acc: 0.5140 - val_loss: 0.9246 - val_acc:		[=============>] - ETA: 15s - loss: 0.8995 - acc: 0.5263
0.5083			[============>] - ETA: 14s - loss: 0.8975 - acc: 0.5275
Epoch ([=====================================
_] - ETA: 16s - loss: 0.8693 - acc: 0.6562		[=====================================
-] - ETA: 15s - loss: 0.8050 - acc: 0.6875	•	[=========>] - ETA: 13s - loss: 0.8998 - acc: 0.5277
] - ETA: 14s - loss: 0.9105 - acc: 0.6250		[=====================================
-	>] - ETA: 13s - loss: 0.9405 - acc: 0.5938		[=====================================
_	>		[=====================================
	>	65/94	[=========>:] - ETA: 12s - loss: 0.8965 - acc: 0.5288
_	=>] - ETA: 12s - loss: 0.9378 - acc: 0.5714		[=========>:] - ETA: 11s - loss: 0.8943 - acc: 0.5308
	=>] - ETA: 12s - loss: 0.9210 - acc: 0.5703		[=====================================
_	=>] - ETA: 12s - loss: 0.9154 - acc: 0.5764		[=====================================
-	[==>] - ETA: 11s - loss: 0.9117 - acc: 0.5563		[=====================================
	[==>] - ETA: 11s - loss: 0.9176 - acc: 0.5483		[=====================================
•	[==>] - ETA: 11s - loss: 0.9094 - acc: 0.5365		[=====================================
_	[===>] - ETA: 11s - loss: 0.9183 - acc: 0.5264		[=====================================
_	[===>] - ETA: 11s - loss: 0.9137 - acc: 0.5246		[=====================================
	[===>] - ETA: 12s - loss: 0.9150 - acc: 0.5292		[=====================================
	[====>] - ETA: 13s - loss: 0.9105 - acc: 0.5352		[=====================================
•	[====>] - ETA: 15s - loss: 0.9114 - acc: 0.5331		[=====================================
	[====>] - ETA: 16s - loss: 0.9014 - acc: 0.5312		[=====================================
	[====>] - ETA: 16s - loss: 0.8955 - acc: 0.5312		[=====================================
-	[====>] - ETA: 17s - loss: 0.8967 - acc: 0.5328		[========0] - ETA: 6s - loss: 0.8899 - acc: 0.5285
Z1/94 [[====>] - ETA: 18s - loss: 0.9097 - acc: 0.5268	80/94	[=====================================

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43/94 [========>.....] - ETA: 19s - loss: 0.8587 - acc: 0.5465
44/94 [========>.....] - ETA: 19s - loss: 0.8641 - acc: 0.5447
45/94 [=======>.....] - ETA: 18s - loss: 0.8661 - acc: 0.5424
46/94 [=======>.....] - ETA: 18s - loss: 0.8672 - acc: 0.5421
47/94 [=======>.....] - ETA: 18s - loss: 0.8658 - acc: 0.5419
48/94 [========>.....] - ETA: 17s - loss: 0.8628 - acc: 0.5449
49/94 [=======>:.....] - ETA: 17s - loss: 0.8638 - acc: 0.5446
50/94 [=======>:.....] - ETA: 17s - loss: 0.8619 - acc: 0.5463
                                                    51/94 [========>:.....] - ETA: 17s - loss: 0.8599 - acc: 0.5472
52/94 [========>:.....] - ETA: 16s - loss: 0.8559 - acc: 0.5493
53/94 [========>:.....] - ETA: 16s - loss: 0.8532 - acc: 0.5495
55/94 [=========>:....] - ETA: 15s - loss: 0.8544 - acc: 0.5500
56/94 [=========>......] - ETA: 15s - loss: 0.8552 - acc: 0.5497
57/94 [=============>.....] - ETA: 15s - loss: 0.8522 - acc: 0.5521
0.5766
                                                    58/94 [=========>:.....] - ETA: 14s - loss: 0.8478 - acc: 0.5550
                                                    59/94 [=========>:.....] - ETA: 14s - loss: 0.8494 - acc: 0.5551
Epoch 7/20
                                                    1/94 [......] - ETA: 16s - loss: 0.6336 - acc: 0.7188
2/94 [......] - ETA: 15s - loss: 0.8330 - acc: 0.5781
                                                    62/94 [==========>:.....] - ETA: 13s - loss: 0.8526 - acc: 0.5534
3/94 [......] - ETA: 14s - loss: 0.8658 - acc: 0.6146
                                                    4/94 [>.....] - ETA: 14s - loss: 0.9146 - acc: 0.5391
5/94 [>......] - ETA: 13s - loss: 0.9345 - acc: 0.5375
                                                    6/94 [>.....] - ETA: 13s - loss: 0.9462 - acc: 0.5365
                                                    7/94 [=>.....] - ETA: 12s - loss: 0.9307 - acc: 0.5268
                                                    66/94 [==========>......] - ETA: 11s - loss: 0.8469 - acc: 0.5540
8/94 [=>.....] - ETA: 12s - loss: 0.9329 - acc: 0.5156
                                                    9/94 [=>.....] - ETA: 12s - loss: 0.9286 - acc: 0.5312
10/94 [==>.....] - ETA: 12s - loss: 0.9009 - acc: 0.5437
                                                    11/94 [==>.....] - ETA: 11s - loss: 0.9035 - acc: 0.5369
                                                    12/94 [==>.....] - ETA: 11s - loss: 0.8968 - acc: 0.5312
                                                    13/94 [===>.....] - ETA: 11s - loss: 0.8845 - acc: 0.5385
14/94 [===>.....] - ETA: 11s - loss: 0.8769 - acc: 0.5446
                                                    15/94 [===>.....] - ETA: 11s - loss: 0.8792 - acc: 0.5417
                                                    16/94 [====>.....] - ETA: 13s - loss: 0.8809 - acc: 0.5371
                                                    17/94 [====>.....] - ETA: 14s - loss: 0.8713 - acc: 0.5423
18/94 [====>.....] - ETA: 15s - loss: 0.8752 - acc: 0.5399
                                                    19/94 [=====>.....] - ETA: 16s - loss: 0.8806 - acc: 0.5312
20/94 [====>.....] - ETA: 17s - loss: 0.8810 - acc: 0.5281
                                                    21/94 [====>.....] - ETA: 18s - loss: 0.8754 - acc: 0.5298
22/94 [====>.....] - ETA: 18s - loss: 0.8761 - acc: 0.5284
                                                    23/94 [=====>.....] - ETA: 19s - loss: 0.8691 - acc: 0.5312
                                                    24/94 [=====>.....] - ETA: 19s - loss: 0.8659 - acc: 0.5326
                                                    25/94 [=====>.....] - ETA: 19s - loss: 0.8709 - acc: 0.5288
                                                    26/94 [======>.....] - ETA: 20s - loss: 0.8698 - acc: 0.5276
                                                    27/94 [======>.....] - ETA: 20s - loss: 0.8798 - acc: 0.5220
                                                    28/94 [======>.....] - ETA: 20s - loss: 0.8844 - acc: 0.5234
29/94 [======>.....] - ETA: 20s - loss: 0.8787 - acc: 0.5291
                                                    30/94 [======>.....] - ETA: 20s - loss: 0.8754 - acc: 0.5302
                                                    31/94 [======>.....] - ETA: 20s - loss: 0.8720 - acc: 0.5312
                                                    32/94 [======>.....] - ETA: 20s - loss: 0.8707 - acc: 0.5293
                                                    33/94 [======>.....] - ETA: 20s - loss: 0.8701 - acc: 0.5284
                                                    34/94 [======>:....] - ETA: 20s - loss: 0.8660 - acc: 0.5312
                                                    35/94 [======>:.....] - ETA: 20s - loss: 0.8638 - acc: 0.5330
                                                    36/94 [=======>......] - ETA: 20s - loss: 0.8635 - acc: 0.5330
37/94 [=======>......] - ETA: 20s - loss: 0.8692 - acc: 0.5346
                                                    0.5536
38/94 [=======>.....] - ETA: 20s - loss: 0.8691 - acc: 0.5378
                                                    Epoch 8/20
39/94 [=======>.....] - ETA: 20s - loss: 0.8669 - acc: 0.5385
                                                    1/94 [.....] - ETA: 15s - loss: 0.9398 - acc: 0.5938
40/94 [=======>......] - ETA: 19s - loss: 0.8636 - acc: 0.5398
                                                    2/94 [......] - ETA: 15s - loss: 0.9462 - acc: 0.5156
41/94 [=======>.....] - ETA: 19s - loss: 0.8604 - acc: 0.5442
                                                    42/94 [========>.....] - ETA: 19s - loss: 0.8595 - acc: 0.5424
                                                    4/94 [>.....] - ETA: 14s - loss: 0.8305 - acc: 0.5859
```

5/94 [>] - ETA: 13s - loss: 0.8329 - acc: 0.5687
6/94 [>] - ETA: 13s - loss: 0.8122 - acc: 0.5573
7/94 [=>] - ETA: 12s - loss: 0.8000 - acc: 0.5804
8/94 [=>] - ETA: 123 loss: 0.0000 acc: 0.5004
9/94 [=>] - ETA: 11s - loss: 0.7819 - acc: 0.5823
10/94 [==>] - ETA: 11s - loss: 0.7858 - acc: 0.5803
11/94 [==>] - ETA: 11s - loss: 0.7636 - acc: 0.3603 11/94 [==>] - ETA: 11s - loss: 0.7791 - acc: 0.5844
12/94 [==>] - ETA: 11s - loss: 0.7845 - acc: 0.5799
13/94 [===>] - ETA: 11s - loss: 0.7823 - acc: 0.5882 14/94 [===>] - ETA: 10s - loss: 0.7811 - acc: 0.5864
15/94 [===>] - ETA: 10s - loss: 0.7611 - acc. 0.5689
16/94 [====>] - ETA: 11s - loss: 0.7762 - acc: 0.5892
17/94 [====>] - ETA: 14s - loss: 0.7702 - acc: 0.5932
18/94 [====>] - ETA: 15s - loss: 0.7820 - acc: 0.6002
19/94 [=====>] - ETA: 16s - loss: 0.7712 - acc: 0.6064
20/94 [====>] - ETA: 175 - loss: 0.7648 - acc: 0.6120
21/94 [====>] - ETA: 18s - loss: 0.7682 - acc: 0.6097
22/94 [=====>] - ETA: 18s - loss: 0.7609 - acc: 0.6104
23/94 [=====>] - ETA: 19s - loss: 0.7675 - acc: 0.6056
24/94 [=====>] - ETA: 19s - loss: 0.7836 - acc: 0.6012
25/94 [=====>] - ETA: 19s - loss: 0.7833 - acc: 0.6021
26/94 [======>] - ETA: 20s - loss: 0.7845 - acc: 0.5970
27/94 [======>] - ETA: 20s - loss: 0.7806 - acc: 0.5946
28/94 [======>] - ETA: 20s - loss: 0.7763 - acc: 0.5968
29/94 [======>] - ETA: 20s - loss: 0.7735 - acc: 0.5967
30/94 [======>] - ETA: 20s - loss: 0.7708 - acc: 0.5966
31/94 [======>] - ETA: 20s - loss: 0.7675 - acc: 0.5965
32/94 [=======>] - ETA: 20s - loss: 0.7733 - acc: 0.5954
33/94 [======>:] - ETA: 20s - loss: 0.7865 - acc: 0.5925
34/94 [=======>] - ETA: 20s - loss: 0.7877 - acc: 0.5916
35/94 [=======>] - ETA: 20s - loss: 0.7867 - acc: 0.5926
36/94 [=======>] - ETA: 20s - loss: 0.7873 - acc: 0.5883
37/94 [=======>] - ETA: 20s - loss: 0.7864 - acc: 0.5910
38/94 [=======>:] - ETA: 20s - loss: 0.7848 - acc: 0.5919
39/94 [========>] - ETA: 20s - loss: 0.7807 - acc: 0.5919
40/94 [=======>:] - ETA: 19s - loss: 0.7787 - acc: 0.5919
41/94 [=======>] - ETA: 19s - loss: 0.7816 - acc: 0.5897
42/94 [=======>] - ETA: 19s - loss: 0.7829 - acc: 0.5876
43/94 [=======>>] - ETA: 19s - loss: 0.7845 - acc: 0.5863
44/94 [========>] - ETA: 19s - loss: 0.7842 - acc: 0.5871
45/94 [=========>] - ETA: 18s - loss: 0.7846 - acc: 0.5852
46/94 [=========>] - ETA: 18s - loss: 0.7850 - acc: 0.5854
47/94 [====================================
48/94 [====================================
49/94 [====================================
51/94 [====================================
52/94 [====================================
53/94 [====================================
54/94 [=======>===>] - ETA: 165 - loss: 0.7744 - acc: 0.5895
55/94 [====================================
56/94 [====================================
57/94 [=======>==>=>] - ETA: 155 - loss: 0.7779 - acc: 0.5881
58/94 [=======>==>=>>] - ETA: 14s - loss: 0.7756 - acc: 0.5871
59/94 [====================================
60/94 [====================================
61/94 [====================================
62/94 [====================================
63/94 [====================================

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64/94 [=========>......] - ETA: 12s - loss: 0.7778 - acc: 0.5858
92/94 [======>:] - ETA: 1s - loss: 0.7724 - acc: 0.5858
0.5831
Epoch 9/20
1/94 [......] - ETA: 16s - loss: 0.6925 - acc: 0.5625
2/94 [......] - ETA: 15s - loss: 0.7230 - acc: 0.6094
3/94 [......] - ETA: 15s - loss: 0.7409 - acc: 0.5625
4/94 [>.....] - ETA: 14s - loss: 0.7451 - acc: 0.5781
5/94 [>......] - ETA: 13s - loss: 0.7328 - acc: 0.5875
6/94 [>.....] - ETA: 13s - loss: 0.7646 - acc: 0.5833
7/94 [=>.....] - ETA: 13s - loss: 0.7831 - acc: 0.5714
8/94 [=>.....] - ETA: 12s - loss: 0.7730 - acc: 0.5781
9/94 [=>.....] - ETA: 12s - loss: 0.7771 - acc: 0.5833
10/94 [==>.....] - ETA: 12s - loss: 0.7606 - acc: 0.5875
11/94 [==>.....] - ETA: 11s - loss: 0.7471 - acc: 0.6040
12/94 [==>.....] - ETA: 11s - loss: 0.7427 - acc: 0.6136
13/94 [===>.....] - ETA: 11s - loss: 0.7434 - acc: 0.6217
14/94 [===>.....] - ETA: 11s - loss: 0.7561 - acc: 0.6264
15/94 [===>.....] - ETA: 12s - loss: 0.7669 - acc: 0.6138
16/94 [====>.....] - ETA: 13s - loss: 0.7730 - acc: 0.6164
17/94 [====>.....] - ETA: 15s - loss: 0.7620 - acc: 0.6316
18/94 [====>.....] - ETA: 16s - loss: 0.7524 - acc: 0.6313
19/94 [====>.....] - ETA: 17s - loss: 0.7611 - acc: 0.6260
20/94 [====>.....] - ETA: 17s - loss: 0.7554 - acc: 0.6275
21/94 [====>.....] - ETA: 18s - loss: 0.7603 - acc: 0.6185
22/94 [====>.....] - ETA: 19s - loss: 0.7611 - acc: 0.6174
23/94 [=====>.....] - ETA: 19s - loss: 0.7673 - acc: 0.6082
24/94 [=====>.....] - ETA: 19s - loss: 0.7664 - acc: 0.6089
25/94 [=====>.....] - ETA: 20s - loss: 0.7680 - acc: 0.6058
```

```
26/94 [=====>.....] - ETA: 20s - loss: 0.7648 - acc: 0.6077
27/94 [======>.....] - ETA: 20s - loss: 0.7623 - acc: 0.6084
28/94 [======>.....] - ETA: 20s - loss: 0.7643 - acc: 0.6101
29/94 [======>.....] - ETA: 20s - loss: 0.7682 - acc: 0.6106
30/94 [======>.....] - ETA: 20s - loss: 0.7663 - acc: 0.6100
31/94 [======>.....] - ETA: 20s - loss: 0.7655 - acc: 0.6085
32/94 [======>.....] - ETA: 20s - loss: 0.7620 - acc: 0.6090
33/94 [=======>.....] - ETA: 20s - loss: 0.7612 - acc: 0.6066
34/94 [======>:....] - ETA: 20s - loss: 0.7662 - acc: 0.6063
35/94 [=======>.....] - ETA: 20s - loss: 0.7712 - acc: 0.6023
36/94 [=======>.....] - ETA: 20s - loss: 0.7757 - acc: 0.6004
37/94 [=======>......] - ETA: 20s - loss: 0.7748 - acc: 0.5976
38/94 [=======>.....] - ETA: 20s - loss: 0.7723 - acc: 0.5967
39/94 [=======>.....] - ETA: 20s - loss: 0.7747 - acc: 0.5966
40/94 [=======>.....] - ETA: 20s - loss: 0.7720 - acc: 0.5966
41/94 [=======>.....] - ETA: 19s - loss: 0.7695 - acc: 0.5988
42/94 [========>.....] - ETA: 19s - loss: 0.7692 - acc: 0.5927
43/94 [========>.....] - ETA: 19s - loss: 0.7700 - acc: 0.5898
44/94 [=======>.....] - ETA: 19s - loss: 0.7674 - acc: 0.5928
45/94 [=========>.....] - ETA: 18s - loss: 0.7672 - acc: 0.5914
46/94 [=======>.....] - ETA: 18s - loss: 0.7675 - acc: 0.5915
47/94 [=======>.....] - ETA: 18s - loss: 0.7661 - acc: 0.5902
48/94 [========>.....] - ETA: 18s - loss: 0.7666 - acc: 0.5909
49/94 [========>.....] - ETA: 17s - loss: 0.7629 - acc: 0.5922
50/94 [=======>.....] - ETA: 17s - loss: 0.7624 - acc: 0.5916
51/94 [=========>......] - ETA: 17s - loss: 0.7596 - acc: 0.5947
52/94 [========>:....] - ETA: 16s - loss: 0.7581 - acc: 0.5965
53/94 [========>......] - ETA: 16s - loss: 0.7575 - acc: 0.5965
55/94 [=========>.....] - ETA: 15s - loss: 0.7563 - acc: 0.5947
56/94 [=========>.....] - ETA: 15s - loss: 0.7553 - acc: 0.5969
57/94 [=========>............] - ETA: 15s - loss: 0.7537 - acc: 0.5968
58/94 [=========>:.....] - ETA: 14s - loss: 0.7505 - acc: 0.5984
59/94 [=========>......] - ETA: 14s - loss: 0.7492 - acc: 0.5978
60/94 [==========>......] - ETA: 14s - loss: 0.7478 - acc: 0.5972
69/94 [==========>.......] - ETA: 10s - loss: 0.7562 - acc: 0.5913
```

```
0.6072
Epoch 10/20
1/94 [......] - ETA: 16s - loss: 0.7006 - acc: 0.7188
2/94 [.....] - ETA: 15s - loss: 0.7390 - acc: 0.6719
3/94 [......] - ETA: 15s - loss: 0.7503 - acc: 0.6562
4/94 [>......] - ETA: 14s - loss: 0.7322 - acc: 0.6719
5/94 [>.....] - ETA: 14s - loss: 0.7379 - acc: 0.6500
6/94 [>......] - ETA: 13s - loss: 0.7310 - acc: 0.6302
7/94 [=>.....] - ETA: 13s - loss: 0.7295 - acc: 0.6250
8/94 [=>.....] - ETA: 12s - loss: 0.7463 - acc: 0.6172
9/94 [=>.....] - ETA: 12s - loss: 0.7632 - acc: 0.6007
10/94 [==>.....] - ETA: 12s - loss: 0.7635 - acc: 0.6062
11/94 [==>.....] - ETA: 12s - loss: 0.7568 - acc: 0.6051
12/94 [==>.....] - ETA: 11s - loss: 0.7582 - acc: 0.6042
13/94 [===>.....] - ETA: 11s - loss: 0.7585 - acc: 0.5938
14/94 [===>.....] - ETA: 11s - loss: 0.7559 - acc: 0.5915
15/94 [===>.....] - ETA: 12s - loss: 0.7480 - acc: 0.5979
16/94 [====>.....] - ETA: 13s - loss: 0.7493 - acc: 0.6035
17/94 [====>.....] - ETA: 14s - loss: 0.7415 - acc: 0.6121
18/94 [====>.....] - ETA: 16s - loss: 0.7387 - acc: 0.6163
19/94 [====>.....] - ETA: 17s - loss: 0.7446 - acc: 0.6201
20/94 [====>.....] - ETA: 17s - loss: 0.7362 - acc: 0.6266
21/94 [====>.....] - ETA: 18s - loss: 0.7296 - acc: 0.6280
22/94 [====>.....] - ETA: 18s - loss: 0.7245 - acc: 0.6307
23/94 [=====>.....] - ETA: 19s - loss: 0.7246 - acc: 0.6277
24/94 [=====>.....] - ETA: 19s - loss: 0.7272 - acc: 0.6250
25/94 [=====>.....] - ETA: 20s - loss: 0.7247 - acc: 0.6275
26/94 [======>.....] - ETA: 20s - loss: 0.7226 - acc: 0.6310
27/94 [======>.....] - ETA: 20s - loss: 0.7270 - acc: 0.6285
28/94 [=====>.....] - ETA: 20s - loss: 0.7276 - acc: 0.6272
29/94 [======>.....] - ETA: 20s - loss: 0.7273 - acc: 0.6239
30/94 [======>.....] - ETA: 21s - loss: 0.7250 - acc: 0.6281
31/94 [======>.....] - ETA: 21s - loss: 0.7273 - acc: 0.6250
32/94 [=======>.....] - ETA: 21s - loss: 0.7245 - acc: 0.6260
33/94 [=======>.....] - ETA: 21s - loss: 0.7208 - acc: 0.6278
34/94 [======>.....] - ETA: 20s - loss: 0.7186 - acc: 0.6259
35/94 [=======>......] - ETA: 20s - loss: 0.7133 - acc: 0.6286
36/94 [=======>.....] - ETA: 20s - loss: 0.7116 - acc: 0.6267
37/94 [=======>.....] - ETA: 20s - loss: 0.7150 - acc: 0.6267
38/94 [=======>.....] - ETA: 20s - loss: 0.7126 - acc: 0.6275
39/94 [=======>.....] - ETA: 20s - loss: 0.7119 - acc: 0.6266
40/94 [======>:.....] - ETA: 20s - loss: 0.7092 - acc: 0.6281
41/94 [========>.....] - ETA: 19s - loss: 0.7076 - acc: 0.6288
42/94 [=======>.....] - ETA: 19s - loss: 0.7101 - acc: 0.6257
43/94 [========>.....] - ETA: 19s - loss: 0.7142 - acc: 0.6265
44/94 [=======>.....] - ETA: 19s - loss: 0.7131 - acc: 0.6264
45/94 [=======>:....] - ETA: 18s - loss: 0.7094 - acc: 0.6271
46/94 [======>:....] - ETA: 18s - loss: 0.7113 - acc: 0.6257
```

47/94 [=	=======>	9/94 [=>] - ETA: 12s - loss: 0.6253 - acc: 0.6597
_	=======>] - ETA: 18s - loss: 0.7097 - acc: 0.6270	_	[==>] - ETA: 12s - loss: 0.6284 - acc: 0.6687
49/94 [=	======>	11/94	[==>] - ETA: 11s - loss: 0.6363 - acc: 0.6705
50/94 [=	======>	12/94	[==>] - ETA: 11s - loss: 0.6376 - acc: 0.6536
51/94 [=	=======>	13/94	[===>] - ETA: 11s - loss: 0.6351 - acc: 0.6587
52/94 [=	:=======>] - ETA: 16s - loss: 0.7127 - acc: 0.6244	14/94	[===>] - ETA: 11s - loss: 0.6402 - acc: 0.6562
	=======>] - ETA: 16s - loss: 0.7103 - acc: 0.6262	15/94	[===>] - ETA: 12s - loss: 0.6530 - acc: 0.6542
	========>] - ETA: 16s - loss: 0.7116 - acc: 0.6256	16/94	[====>] - ETA: 13s - loss: 0.6514 - acc: 0.6621
	========>] - ETA: 15s - loss: 0.7150 - acc: 0.6233	17/94	[====>] - ETA: 14s - loss: 0.6530 - acc: 0.6562
56/94 [=	========>] - ETA: 15s - loss: 0.7121 - acc: 0.6267	18/94	[====>] - ETA: 16s - loss: 0.6589 - acc: 0.6562
57/94 [=	========>] - ETA: 15s - loss: 0.7131 - acc: 0.6255	19/94	[====>] - ETA: 16s - loss: 0.6594 - acc: 0.6546
58/94 [=	=======>] - ETA: 14s - loss: 0.7143 - acc: 0.6250	20/94	[====>] - ETA: 17s - loss: 0.6690 - acc: 0.6500
59/94 [=	========>] - ETA: 14s - loss: 0.7141 - acc: 0.6261	21/94	[====>] - ETA: 18s - loss: 0.6715 - acc: 0.6503
60/94 [=	========>] - ETA: 14s - loss: 0.7155 - acc: 0.6240	22/94	[====>] - ETA: 18s - loss: 0.6729 - acc: 0.6463
	======================================	23/94	[=====>] - ETA: 19s - loss: 0.6705 - acc: 0.6427
62/94 [=	========>] - ETA: 13s - loss: 0.7152 - acc: 0.6220	24/94	[=====>] - ETA: 19s - loss: 0.6641 - acc: 0.6445
63/94 [=	======================================	25/94	[=====>] - ETA: 19s - loss: 0.6641 - acc: 0.6412
	========>:] - ETA: 12s - loss: 0.7132 - acc: 0.6230	26/94	[======>] - ETA: 20s - loss: 0.6664 - acc: 0.6394
	========>:] - ETA: 12s - loss: 0.7127 - acc: 0.6221	27/94	[======>] - ETA: 20s - loss: 0.6702 - acc: 0.6366
	=======>] - ETA: 11s - loss: 0.7133 - acc: 0.6207	28/94	[======>] - ETA: 20s - loss: 0.6775 - acc: 0.6384
67/94 [=	=========>] - ETA: 11s - loss: 0.7146 - acc: 0.6213		[======>] - ETA: 20s - loss: 0.6832 - acc: 0.6379
	=========>] - ETA: 11s - loss: 0.7155 - acc: 0.6204	30/94	[======>] - ETA: 20s - loss: 0.6852 - acc: 0.6385
69/94 [=	=========>] - ETA: 10s - loss: 0.7169 - acc: 0.6200		[======>] - ETA: 20s - loss: 0.6835 - acc: 0.6411
	======================================	32/94	[======>] - ETA: 20s - loss: 0.6849 - acc: 0.6387
71/94 [=	======================================		[======>] - ETA: 20s - loss: 0.6859 - acc: 0.6383
72/94 [=	======================================	34/94	[======>] - ETA: 20s - loss: 0.6841 - acc: 0.6379
73/94 [=	======================================	35/94	[=======>] - ETA: 20s - loss: 0.6869 - acc: 0.6384
74/94 [=	======================================	36/94	[======>] - ETA: 20s - loss: 0.6863 - acc: 0.6415
75/94 [=	=========>============================	37/94	[======>:] - ETA: 20s - loss: 0.6842 - acc: 0.6419
	======================================	38/94	[=======>] - ETA: 20s - loss: 0.6835 - acc: 0.6402
	======================================	39/94	[=======>] - ETA: 19s - loss: 0.6798 - acc: 0.6414
	======================================	40/94	[=======>] - ETA: 19s - loss: 0.6781 - acc: 0.6402
79/94 [=	======================================	41/94	[=======>] - ETA: 19s - loss: 0.6774 - acc: 0.6436
	========>=============================	42/94	[=======>] - ETA: 19s - loss: 0.6767 - acc: 0.6447
81/94 [=	=========>] - ETA: 5s - loss: 0.7120 - acc: 0.6206	43/94	[======>:] - ETA: 19s - loss: 0.6799 - acc: 0.6435
82/94 [=	==========>] - ETA: 5s - loss: 0.7089 - acc: 0.6215	44/94	[=======>:] - ETA: 18s - loss: 0.6786 - acc: 0.6438
83/94 [=	======================================	45/94	[========>] - ETA: 18s - loss: 0.6793 - acc: 0.6406
84/94 [=	======================================		[======>:] - ETA: 18s - loss: 0.6783 - acc: 0.6396
85/94 [=	======================================	47/94	[======>:] - ETA: 18s - loss: 0.6793 - acc: 0.6379
86/94 [=	======================================		[=======>:] - ETA: 17s - loss: 0.6802 - acc: 0.6383
87/94 [=	======================================	49/94	[=======>:] - ETA: 17s - loss: 0.6813 - acc: 0.6387
88/94 [=	======================================	50/94	[=======>:] - ETA: 17s - loss: 0.6832 - acc: 0.6397
89/94 [=	======================================	51/94	[============>] - ETA: 16s - loss: 0.6885 - acc: 0.6382
90/94 [=			[======>:] - ETA: 16s - loss: 0.6889 - acc: 0.6361
		53/94	[======>:] - ETA: 16s - loss: 0.6890 - acc: 0.6353
92/94 [=	>.] - ETA: 1s - loss: 0.7057 - acc: 0.6235	54/94	[=====================================
	======================================	55/94	[========>:] - ETA: 15s - loss: 0.6913 - acc: 0.6349
94/94 [=	======================================		[=======>:] - ETA: 15s - loss: 0.6902 - acc: 0.6336
95/94 [=	======================================	57/94	[=====================================
0.6261		58/94	[=====================================
Epoch 1:	1/20	59/94	[=====================================
1/94 [] - ETA: 16s - loss: 0.6236 - acc: 0.6250	60/94	[=====================================
2/94 [] - ETA: 15s - loss: 0.5878 - acc: 0.6406		[=====================================
3/94 [] - ETA: 14s - loss: 0.6668 - acc: 0.6250	62/94	[=====================================
4/94 [>.] - ETA: 14s - loss: 0.6746 - acc: 0.6250		[=====================================
5/94 [>.] - ETA: 13s - loss: 0.6631 - acc: 0.6188	64/94	[=====================================
6/94 [>.] - ETA: 13s - loss: 0.6572 - acc: 0.6302	65/94	[============>] - ETA: 12s - loss: 0.6950 - acc: 0.6252
	>] - ETA: 13s - loss: 0.6411 - acc: 0.6250	66/94	[=====================================
8/94 [=:	>] - ETA: 12s - loss: 0.6337 - acc: 0.6367		[=====================================

68/94 [=	:=====================================	30/94 [==	=====>] - ETA: 21s - loss: 0.6614 - acc: 0.6469
_	=======>:] - ETA: 10s - loss: 0.6892 - acc: 0.6252	=	=====>>] - ETA: 21s - loss: 0.6554 - acc: 0.6492
_	======================================	_	======>
-	======================================		======>] - ETA: 21s - loss: 0.6597 - acc: 0.6449
_	======================================	=	======>] - ETA: 20s - loss: 0.6604 - acc: 0.6443
_	======================================	_	======>] - ETA: 20s - loss: 0.6643 - acc: 0.6455
-	======================================		======>] - ETA: 20s - loss: 0.6660 - acc: 0.6441
	======================================		======>
_	======================================		=======>] - ETA: 20s - loss: 0.6641 - acc: 0.6488
	======================================		=======>] - ETA: 20s - loss: 0.6643 - acc: 0.6474
	======================================		=======>:] - ETA: 20s - loss: 0.6629 - acc: 0.6500
-	•		=======>] - ETA: 20s - loss: 0.6601 - acc: 0.6494
		41/94 [==	
		42/94 [==	========>] - ETA: 19s - loss: 0.6603 - acc: 0.6458
_			=======>] - ETA: 19s - loss: 0.6635 - acc: 0.6424
-			=======>] - ETA: 19s - loss: 0.6628 - acc: 0.6428
_			=======>] - ETA: 18s - loss: 0.6610 - acc: 0.6417
	======================================	_	======>
	>] - ETA: 4s - loss: 0.6918 - acc: 0.6237	=	=======>] - ETA: 18s - loss: 0.6617 - acc: 0.6426
	======================================	=	=======>] - ETA: 18s - loss: 0.6629 - acc: 0.6409
	======================================	_	=======>] - ETA: 17s - loss: 0.6645 - acc: 0.6393
_	======================================		========>] - ETA: 17s - loss: 0.6670 - acc: 0.6390
	======================================		=======>] - ETA: 17s - loss: 0.6646 - acc: 0.6431
	======================================		========>] - ETA: 16s - loss: 0.6642 - acc: 0.6433
	======================================		=======>:] - ETA: 16s - loss: 0.6644 - acc: 0.6453
_		-	=======>:] - ETA: 16s - loss: 0.6696 - acc: 0.6420
93/94 [=	======================================	55/94 [==	========>] - ETA: 15s - loss: 0.6699 - acc: 0.6417
94/94 [=		56/94 [==	========>] - ETA: 15s - loss: 0.6714 - acc: 0.6420
95/94 [=	======================================	57/94 [==	=========>] - ETA: 15s - loss: 0.6724 - acc: 0.6406
0.6340			=========>] - ETA: 14s - loss: 0.6708 - acc: 0.6414
Epoch 12	2/20	59/94 [==	========>] - ETA: 14s - loss: 0.6697 - acc: 0.6406
1/94 [] - ETA: 15s - loss: 0.6703 - acc: 0.5000		=========>] - ETA: 14s - loss: 0.6709 - acc: 0.6398
2/94 [] - ETA: 15s - loss: 0.5116 - acc: 0.7031	61/94 [==	========>:] - ETA: 13s - loss: 0.6693 - acc: 0.6416
3/94 [] - ETA: 14s - loss: 0.5416 - acc: 0.6667		========>:] - ETA: 13s - loss: 0.6691 - acc: 0.6419
4/94 [>.] - ETA: 13s - loss: 0.5488 - acc: 0.6641	63/94 [==	=========>] - ETA: 13s - loss: 0.6674 - acc: 0.6446
5/94 [>.] - ETA: 13s - loss: 0.5682 - acc: 0.6875	64/94 [==	========>:] - ETA: 12s - loss: 0.6666 - acc: 0.6448
6/94 [>.] - ETA: 12s - loss: 0.5836 - acc: 0.6823		======================================
7/94 [=>	>] - ETA: 12s - loss: 0.5846 - acc: 0.6741		======================================
8/94 [=>	>] - ETA: 12s - loss: 0.6004 - acc: 0.6836		========>] - ETA: 11s - loss: 0.6658 - acc: 0.6429
=	>] - ETA: 12s - loss: 0.5885 - acc: 0.7049		======================================
_	=>] - ETA: 11s - loss: 0.6108 - acc: 0.6844	69/94 [==	======================================
11/94 [=	=>] - ETA: 11s - loss: 0.6245 - acc: 0.6705		=========>] - ETA: 10s - loss: 0.6639 - acc: 0.6466
	=>] - ETA: 11s - loss: 0.6259 - acc: 0.6693		======================================
13/94 [=	:==>] - ETA: 11s - loss: 0.6284 - acc: 0.6707	72/94 [==	=========>:] - ETA: 9s - loss: 0.6634 - acc: 0.6456
_	:==>] - ETA: 11s - loss: 0.6332 - acc: 0.6652		======================================
_	==>] - ETA: 12s - loss: 0.6220 - acc: 0.6771		======================================
_	===>	-	======================================
	===>] - ETA: 14s - loss: 0.6230 - acc: 0.6783	-	======================================
	===>		======================================
-	====>] - ETA: 16s - loss: 0.6380 - acc: 0.6694		======================================
_	====>] - ETA: 17s - loss: 0.6443 - acc: 0.6641	79/94 [==	======================================
_			======================================
_	:====> 1 - FTA: 18s - loss: 0.6468 - acc: 0.6548	80/94 (==	
22/94 l=	:===>	=	-
-	====>] - ETA: 18s - loss: 0.6446 - acc: 0.6562	81/94 [==	======================================
23/94 [=	====>] - ETA: 18s - loss: 0.6446 - acc: 0.6562 =====>] - ETA: 19s - loss: 0.6525 - acc: 0.6454	81/94 [== 82/94 [==	======================================
23/94 [= 24/94 [=	====>] - ETA: 18s - loss: 0.6446 - acc: 0.6562 =====>] - ETA: 19s - loss: 0.6525 - acc: 0.6454 =====>] - ETA: 19s - loss: 0.6589 - acc: 0.6445	81/94 [== 82/94 [== 83/94 [==	======================================
23/94 [= 24/94 [= 25/94 [=	====>	81/94 [== 82/94 [== 83/94 [== 84/94 [==	======================================
23/94 [= 24/94 [= 25/94 [= 26/94 [=	====>	81/94 [====================================	======================================
23/94 [= 24/94 [= 25/94 [= 26/94 [= 27/94 [=	====>	81/94 [====================================	======================================
23/94 [= 24/94 [= 25/94 [= 26/94 [= 27/94 [= 28/94 [=	====>	81/94 [====================================	======================================

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51/94 [========>:.....] - ETA: 16s - loss: 0.6236 - acc: 0.6465
52/94 [========>:....] - ETA: 16s - loss: 0.6248 - acc: 0.6461
                                                               53/94 [==============>.....] - ETA: 16s - loss: 0.6275 - acc: 0.6451
54/94 [=========>......] - ETA: 16s - loss: 0.6256 - acc: 0.6471
55/94 [=========>......] - ETA: 15s - loss: 0.6259 - acc: 0.6478
56/94 [=========>.....] - ETA: 15s - loss: 0.6255 - acc: 0.6474
58/94 [=========>:.....] - ETA: 14s - loss: 0.6267 - acc: 0.6483
0.6422
                                                               59/94 [==========>.....] - ETA: 14s - loss: 0.6306 - acc: 0.6457
Epoch 13/20
1/94 [......] - ETA: 16s - loss: 0.4698 - acc: 0.8438
                                                               60/94 [=========>:.....] - ETA: 14s - loss: 0.6277 - acc: 0.6470
                                                               2/94 [......] - ETA: 14s - loss: 0.5291 - acc: 0.7680
                                                               62/94 [==============>......] - ETA: 13s - loss: 0.6292 - acc: 0.6447
3/94 [......] - ETA: 13s - loss: 0.5527 - acc: 0.7099
                                                               63/94 [==========>:.........] - ETA: 12s - loss: 0.6300 - acc: 0.6434
4/94 [>.....] - ETA: 13s - loss: 0.5585 - acc: 0.7121
5/94 [>......] - ETA: 13s - loss: 0.5838 - acc: 0.6822
                                                               6/94 [>.....] - ETA: 12s - loss: 0.6014 - acc: 0.6675
                                                               65/94 [=========>......] - ETA: 12s - loss: 0.6338 - acc: 0.6429
7/94 [=>.....] - ETA: 12s - loss: 0.5991 - acc: 0.6659
                                                               66/94 [=========>......] - ETA: 11s - loss: 0.6350 - acc: 0.6426
                                                               8/94 [=>.....] - ETA: 12s - loss: 0.6054 - acc: 0.6569
9/94 [=>.....] - ETA: 12s - loss: 0.5964 - acc: 0.6637
                                                               10/94 [==>.....] - ETA: 11s - loss: 0.6148 - acc: 0.6692
                                                               11/94 [==>.....] - ETA: 11s - loss: 0.6167 - acc: 0.6766
12/94 [==>.....] - ETA: 11s - loss: 0.6298 - acc: 0.6697
                                                               13/94 [===>.....] - ETA: 11s - loss: 0.6278 - acc: 0.6734
                                                               14/94 [===>.....] - ETA: 11s - loss: 0.6281 - acc: 0.6700
                                                               15/94 [===>.....] - ETA: 11s - loss: 0.6241 - acc: 0.6691
                                                               16/94 [====>.....] - ETA: 13s - loss: 0.6197 - acc: 0.6702
                                                               17/94 [====>.....] - ETA: 14s - loss: 0.6212 - acc: 0.6694
18/94 [====>.....] - ETA: 15s - loss: 0.6276 - acc: 0.6669
                                                               19/94 [====>.....] - ETA: 16s - loss: 0.6225 - acc: 0.6631
                                                               20/94 [====>.....] - ETA: 17s - loss: 0.6233 - acc: 0.6534
                                                               21/94 [====>.....] - ETA: 18s - loss: 0.6351 - acc: 0.6520
22/94 [====>.....] - ETA: 18s - loss: 0.6386 - acc: 0.6508
                                                               23/94 [=====>.....] - ETA: 19s - loss: 0.6299 - acc: 0.6578
                                                               24/94 [=====>.....] - ETA: 19s - loss: 0.6259 - acc: 0.6565
                                                               25/94 [=====>.....] - ETA: 19s - loss: 0.6259 - acc: 0.6514
                                                               84/94 [=======>:...] - ETA: 4s - loss: 0.6365 - acc: 0.6385
                                                               26/94 [======>.....] - ETA: 20s - loss: 0.6348 - acc: 0.6504
                                                               27/94 [======>.....] - ETA: 20s - loss: 0.6333 - acc: 0.6472
28/94 [======>.....] - ETA: 20s - loss: 0.6377 - acc: 0.6441
                                                               29/94 [======>.....] - ETA: 20s - loss: 0.6400 - acc: 0.6435
30/94 [======>.....] - ETA: 20s - loss: 0.6350 - acc: 0.6439
                                                               31/94 [======>.....] - ETA: 20s - loss: 0.6318 - acc: 0.6453
                                                               32/94 [======>.....] - ETA: 20s - loss: 0.6323 - acc: 0.6447
                                                               33/94 [======>:.....] - ETA: 20s - loss: 0.6318 - acc: 0.6469
                                                               34/94 [======>:....] - ETA: 20s - loss: 0.6387 - acc: 0.6435
                                                               35/94 [=======>......] - ETA: 20s - loss: 0.6414 - acc: 0.6403
36/94 [=======>.....] - ETA: 20s - loss: 0.6412 - acc: 0.6425
                                                               37/94 [=======>......] - ETA: 20s - loss: 0.6411 - acc: 0.6437
                                                               0.6561
38/94 [=======>.....] - ETA: 20s - loss: 0.6379 - acc: 0.6457
                                                               Epoch 14/20
39/94 [=======>.....] - ETA: 19s - loss: 0.6390 - acc: 0.6452
                                                               1/94 [......] - ETA: 16s - loss: 0.4381 - acc: 0.7500
40/94 [=======>.....] - ETA: 19s - loss: 0.6357 - acc: 0.6447
                                                               41/94 [=======>.....] - ETA: 19s - loss: 0.6352 - acc: 0.6457
                                                               3/94 [.....] - ETA: 14s - loss: 0.5752 - acc: 0.6875
42/94 [=======>:....] - ETA: 19s - loss: 0.6327 - acc: 0.6474
                                                               4/94 [>.....] - ETA: 14s - loss: 0.6128 - acc: 0.6641
43/94 [========>.....] - ETA: 19s - loss: 0.6311 - acc: 0.6491
                                                               5/94 [>.....] - ETA: 13s - loss: 0.5686 - acc: 0.7000
44/94 [========>......] - ETA: 18s - loss: 0.6289 - acc: 0.6493
                                                               6/94 [>.....] - ETA: 13s - loss: 0.5699 - acc: 0.6979
45/94 [=======>.....] - ETA: 18s - loss: 0.6286 - acc: 0.6480
                                                               7/94 [=>.....] - ETA: 12s - loss: 0.5729 - acc: 0.6875
46/94 [=======>.....] - ETA: 18s - loss: 0.6267 - acc: 0.6482
                                                               8/94 [=>.....] - ETA: 12s - loss: 0.5742 - acc: 0.6797
47/94 [========>.....] - ETA: 18s - loss: 0.6263 - acc: 0.6477
                                                               9/94 [=>.....] - ETA: 12s - loss: 0.5650 - acc: 0.6806
48/94 [=======>:.....] - ETA: 17s - loss: 0.6227 - acc: 0.6505
                                                               10/94 [==>.....] - ETA: 11s - loss: 0.5618 - acc: 0.6781
49/94 [=======>.....] - ETA: 17s - loss: 0.6244 - acc: 0.6493
                                                               11/94 [==>.....] - ETA: 11s - loss: 0.5620 - acc: 0.6705
50/94 [========>.....] - ETA: 17s - loss: 0.6257 - acc: 0.6470
                                                               12/94 [==>.....] - ETA: 11s - loss: 0.5618 - acc: 0.6771
```

```
13/94 [===>.....] - ETA: 11s - loss: 0.5630 - acc: 0.6755
14/94 [===>.....] - ETA: 11s - loss: 0.5708 - acc: 0.6719
15/94 [===>.....] - ETA: 11s - loss: 0.5636 - acc: 0.6708
16/94 [====>.....] - ETA: 13s - loss: 0.5683 - acc: 0.6758
17/94 [====>.....] - ETA: 14s - loss: 0.5763 - acc: 0.6765
18/94 [====>.....] - ETA: 15s - loss: 0.5770 - acc: 0.6771
19/94 [====>.....] - ETA: 16s - loss: 0.5730 - acc: 0.6809
20/94 [====>.....] - ETA: 17s - loss: 0.5780 - acc: 0.6703
21/94 [====>.....] - ETA: 18s - loss: 0.5740 - acc: 0.6771
22/94 [====>.....] - ETA: 18s - loss: 0.5732 - acc: 0.6804
23/94 [=====>.....] - ETA: 19s - loss: 0.5776 - acc: 0.6753
24/94 [=====>.....] - ETA: 19s - loss: 0.5808 - acc: 0.6760
25/94 [=====>.....] - ETA: 19s - loss: 0.5781 - acc: 0.6777
26/94 [======>.....] - ETA: 19s - loss: 0.5749 - acc: 0.6781
27/94 [=====>.....] - ETA: 20s - loss: 0.5749 - acc: 0.6773
28/94 [======>.....] - ETA: 20s - loss: 0.5754 - acc: 0.6799
29/94 [======>.....] - ETA: 20s - loss: 0.5765 - acc: 0.6780
30/94 [======>.....] - ETA: 20s - loss: 0.5775 - acc: 0.6731
31/94 [======>.....] - ETA: 20s - loss: 0.5827 - acc: 0.6695
32/94 [=======>.....] - ETA: 20s - loss: 0.6007 - acc: 0.6632
33/94 [=======>.....] - ETA: 20s - loss: 0.6045 - acc: 0.6592
34/94 [=======>.....] - ETA: 20s - loss: 0.6021 - acc: 0.6582
35/94 [=======>.....] - ETA: 20s - loss: 0.6028 - acc: 0.6537
36/94 [=======>.....] - ETA: 20s - loss: 0.6041 - acc: 0.6555
37/94 [=======>.....] - ETA: 20s - loss: 0.6088 - acc: 0.6547
38/94 [========>.....] - ETA: 19s - loss: 0.6098 - acc: 0.6506
39/94 [=======>.....] - ETA: 19s - loss: 0.6139 - acc: 0.6516
40/94 [=======>.....] - ETA: 19s - loss: 0.6155 - acc: 0.6486
41/94 [========>.....] - ETA: 19s - loss: 0.6143 - acc: 0.6487
42/94 [=======>.....] - ETA: 19s - loss: 0.6151 - acc: 0.6504
43/94 [=======>.....] - ETA: 19s - loss: 0.6129 - acc: 0.6513
44/94 [=======>.....] - ETA: 18s - loss: 0.6166 - acc: 0.6507
45/94 [========>.....] - ETA: 18s - loss: 0.6192 - acc: 0.6487
46/94 [=======>.....] - ETA: 18s - loss: 0.6201 - acc: 0.6468
47/94 [=======>.....] - ETA: 18s - loss: 0.6212 - acc: 0.6450
48/94 [=======>:....] - ETA: 17s - loss: 0.6180 - acc: 0.6505
49/94 [=======>.....] - ETA: 17s - loss: 0.6178 - acc: 0.6500
50/94 [========>.....] - ETA: 17s - loss: 0.6170 - acc: 0.6513
51/94 [========>......] - ETA: 16s - loss: 0.6166 - acc: 0.6508
52/94 [========>......] - ETA: 16s - loss: 0.6151 - acc: 0.6533
53/94 [========>:....] - ETA: 16s - loss: 0.6158 - acc: 0.6528
54/94 [=========>.....] - ETA: 16s - loss: 0.6143 - acc: 0.6534
55/94 [=========>.....] - ETA: 15s - loss: 0.6138 - acc: 0.6529
57/94 [=========>......] - ETA: 15s - loss: 0.6141 - acc: 0.6498
58/94 [=========>......] - ETA: 14s - loss: 0.6179 - acc: 0.6499
59/94 [=========>......] - ETA: 14s - loss: 0.6255 - acc: 0.6489
64/94 [=========>......] - ETA: 12s - loss: 0.6235 - acc: 0.6510
66/94 [======>:.....] - ETA: 11s - loss: 0.6220 - acc: 0.6506
69/94 [========>:......] - ETA: 10s - loss: 0.6286 - acc: 0.6495
```

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84/94 [======>:...] - ETA: 4s - loss: 0.6304 - acc: 0.6518
0.6447
Epoch 15/20
1/94 [......] - ETA: 15s - loss: 0.4758 - acc: 0.6875
2/94 [......] - ETA: 15s - loss: 0.4924 - acc: 0.7031
3/94 [......] - ETA: 14s - loss: 0.5832 - acc: 0.6354
4/94 [>.....] - ETA: 14s - loss: 0.6564 - acc: 0.6094
5/94 [>......] - ETA: 13s - loss: 0.6530 - acc: 0.6375
6/94 [>.....] - ETA: 13s - loss: 0.6490 - acc: 0.6406
7/94 [=>.....] - ETA: 12s - loss: 0.6686 - acc: 0.6518
8/94 [=>.....] - ETA: 12s - loss: 0.6710 - acc: 0.6484
9/94 [=>.....] - ETA: 12s - loss: 0.6636 - acc: 0.6424
10/94 [==>.....] - ETA: 11s - loss: 0.6440 - acc: 0.6531
11/94 [==>.....] - ETA: 11s - loss: 0.6314 - acc: 0.6562
12/94 [==>.....] - ETA: 11s - loss: 0.6263 - acc: 0.6641
13/94 [===>.....] - ETA: 11s - loss: 0.6180 - acc: 0.6611
14/94 [===>.....] - ETA: 11s - loss: 0.6103 - acc: 0.6652
15/94 [===>.....] - ETA: 12s - loss: 0.6065 - acc: 0.6667
16/94 [====>.....] - ETA: 13s - loss: 0.5967 - acc: 0.6680
17/94 [====>.....] - ETA: 14s - loss: 0.5941 - acc: 0.6691
18/94 [====>.....] - ETA: 15s - loss: 0.6087 - acc: 0.6684
19/94 [====>.....] - ETA: 16s - loss: 0.6103 - acc: 0.6628
20/94 [====>.....] - ETA: 17s - loss: 0.6101 - acc: 0.6656
21/94 [====>.....] - ETA: 18s - loss: 0.6036 - acc: 0.6682
22/94 [====>.....] - ETA: 18s - loss: 0.6012 - acc: 0.6719
23/94 [=====>.....] - ETA: 19s - loss: 0.5950 - acc: 0.6793
24/94 [=====>.....] - ETA: 19s - loss: 0.5970 - acc: 0.6719
25/94 [=====>.....] - ETA: 19s - loss: 0.5958 - acc: 0.6700
26/94 [======>.....] - ETA: 20s - loss: 0.5944 - acc: 0.6695
27/94 [======>.....] - ETA: 20s - loss: 0.6003 - acc: 0.6678
28/94 [======>.....] - ETA: 20s - loss: 0.5981 - acc: 0.6730
29/94 [======>.....] - ETA: 20s - loss: 0.5959 - acc: 0.6724
30/94 [======>.....] - ETA: 20s - loss: 0.5919 - acc: 0.6760
31/94 [======>.....] - ETA: 20s - loss: 0.5917 - acc: 0.6744
32/94 [======>.....] - ETA: 20s - loss: 0.5885 - acc: 0.6768
33/94 [=======>.....] - ETA: 20s - loss: 0.5871 - acc: 0.6761
```

```
34/94 [======>:....] - ETA: 20s - loss: 0.5809 - acc: 0.6792
35/94 [=======>......] - ETA: 20s - loss: 0.5837 - acc: 0.6750
36/94 [=======>......] - ETA: 20s - loss: 0.5843 - acc: 0.6701
37/94 [=======>.....] - ETA: 20s - loss: 0.5821 - acc: 0.6706
38/94 [=======>.....] - ETA: 19s - loss: 0.5822 - acc: 0.6732
39/94 [=======>.....] - ETA: 19s - loss: 0.5802 - acc: 0.6744
40/94 [=======>.....] - ETA: 19s - loss: 0.5794 - acc: 0.6731
41/94 [=======>.....] - ETA: 19s - loss: 0.5810 - acc: 0.6689
42/94 [========>.....] - ETA: 19s - loss: 0.5808 - acc: 0.6679
43/94 [========>.....] - ETA: 18s - loss: 0.5813 - acc: 0.6676
44/94 [========>......] - ETA: 18s - loss: 0.5820 - acc: 0.6659
45/94 [=========>.....] - ETA: 18s - loss: 0.5815 - acc: 0.6699
46/94 [=======>.....] - ETA: 18s - loss: 0.5787 - acc: 0.6709
47/94 [=======>.....] - ETA: 17s - loss: 0.5812 - acc: 0.6713
48/94 [========>.....] - ETA: 17s - loss: 0.5794 - acc: 0.6716
49/94 [=======>:....] - ETA: 17s - loss: 0.5840 - acc: 0.6707
50/94 [========>.....] - ETA: 17s - loss: 0.5858 - acc: 0.6685
51/94 [========>:....] - ETA: 16s - loss: 0.5849 - acc: 0.6658
52/94 [========>.....] - ETA: 16s - loss: 0.5888 - acc: 0.6626
55/94 [=========>.....] - ETA: 15s - loss: 0.5912 - acc: 0.6583
56/94 [=========>.....] - ETA: 15s - loss: 0.5898 - acc: 0.6566
57/94 [=========>:.....] - ETA: 14s - loss: 0.5957 - acc: 0.6533
58/94 [==========>......] - ETA: 14s - loss: 0.6022 - acc: 0.6496
59/94 [=========>......] - ETA: 14s - loss: 0.6101 - acc: 0.6492
60/94 [=========>.....] - ETA: 13s - loss: 0.6105 - acc: 0.6498
66/94 [=========>......] - ETA: 11s - loss: 0.6023 - acc: 0.6546
68/94 [==========>:.......] - ETA: 11s - loss: 0.6031 - acc: 0.6533
```

```
0.6541
Epoch 16/20
1/94 [.....] - ETA: 16s - loss: 0.4797 - acc: 0.8438
2/94 [......] - ETA: 15s - loss: 0.5854 - acc: 0.7031
3/94 [......] - ETA: 14s - loss: 0.5684 - acc: 0.7083
4/94 [>.....] - ETA: 14s - loss: 0.5637 - acc: 0.6953
5/94 [>.....] - ETA: 13s - loss: 0.6158 - acc: 0.6687
6/94 [>.....] - ETA: 13s - loss: 0.6152 - acc: 0.6615
7/94 [=>.....] - ETA: 12s - loss: 0.6262 - acc: 0.6562
8/94 [=>.....] - ETA: 12s - loss: 0.6247 - acc: 0.6523
9/94 [=>.....] - ETA: 12s - loss: 0.6230 - acc: 0.6528
10/94 [==>.....] - ETA: 12s - loss: 0.6120 - acc: 0.6531
11/94 [==>.....] - ETA: 11s - loss: 0.6145 - acc: 0.6477
12/94 [==>.....] - ETA: 11s - loss: 0.6042 - acc: 0.6589
13/94 [===>.....] - ETA: 11s - loss: 0.6206 - acc: 0.6538
14/94 [===>.....] - ETA: 11s - loss: 0.6296 - acc: 0.6473
15/94 [===>.....] - ETA: 11s - loss: 0.6269 - acc: 0.6396
16/94 [====>.....] - ETA: 13s - loss: 0.6258 - acc: 0.6367
17/94 [====>.....] - ETA: 14s - loss: 0.6279 - acc: 0.6379
18/94 [====>.....] - ETA: 15s - loss: 0.6179 - acc: 0.6441
19/94 [====>.....] - ETA: 16s - loss: 0.6137 - acc: 0.6447
20/94 [====>.....] - ETA: 17s - loss: 0.6120 - acc: 0.6500
21/94 [====>.....] - ETA: 18s - loss: 0.6133 - acc: 0.6443
22/94 [====>.....] - ETA: 18s - loss: 0.6195 - acc: 0.6435
23/94 [=====>.....] - ETA: 19s - loss: 0.6216 - acc: 0.6508
24/94 [=====>.....] - ETA: 19s - loss: 0.6265 - acc: 0.6484
25/94 [=====>.....] - ETA: 20s - loss: 0.6261 - acc: 0.6425
26/94 [=====>.....] - ETA: 20s - loss: 0.6289 - acc: 0.6406
27/94 [======>.....] - ETA: 20s - loss: 0.6274 - acc: 0.6435
28/94 [======>.....] - ETA: 20s - loss: 0.6243 - acc: 0.6484
29/94 [======>.....] - ETA: 20s - loss: 0.6212 - acc: 0.6519
30/94 [======>......] - ETA: 20s - loss: 0.6248 - acc: 0.6490
31/94 [=======>.....] - ETA: 20s - loss: 0.6224 - acc: 0.6512
32/94 [======>.....] - ETA: 20s - loss: 0.6195 - acc: 0.6514
33/94 [=======>.....] - ETA: 20s - loss: 0.6195 - acc: 0.6515
34/94 [=======>.....] - ETA: 20s - loss: 0.6185 - acc: 0.6498
35/94 [=======>.....] - ETA: 20s - loss: 0.6160 - acc: 0.6518
36/94 [=======>.....] - ETA: 20s - loss: 0.6117 - acc: 0.6554
37/94 [=======>.....] - ETA: 20s - loss: 0.6122 - acc: 0.6537
38/94 [=======>:.....] - ETA: 20s - loss: 0.6159 - acc: 0.6521
39/94 [========>.....] - ETA: 20s - loss: 0.6155 - acc: 0.6506
40/94 [========>.....] - ETA: 19s - loss: 0.6135 - acc: 0.6523
41/94 [======>:....] - ETA: 19s - loss: 0.6102 - acc: 0.6540
42/94 [========>.....] - ETA: 19s - loss: 0.6067 - acc: 0.6548
43/94 [=======>.....] - ETA: 19s - loss: 0.6036 - acc: 0.6570
44/94 [=======>.....] - ETA: 19s - loss: 0.6020 - acc: 0.6577
45/94 [=======>.....] - ETA: 18s - loss: 0.5986 - acc: 0.6590
46/94 [=======>.....] - ETA: 18s - loss: 0.6025 - acc: 0.6576
47/94 [========>.....] - ETA: 18s - loss: 0.6031 - acc: 0.6562
48/94 [========>.....] - ETA: 18s - loss: 0.5996 - acc: 0.6602
49/94 [=======>:.....] - ETA: 17s - loss: 0.5979 - acc: 0.6607
50/94 [========>.....] - ETA: 17s - loss: 0.5951 - acc: 0.6613
51/94 [=======>:....] - ETA: 17s - loss: 0.5935 - acc: 0.6630
52/94 [=========>......] - ETA: 16s - loss: 0.5930 - acc: 0.6653
53/94 [==============>.....] - ETA: 16s - loss: 0.5950 - acc: 0.6651
54/94 [=========>......] - ETA: 16s - loss: 0.5939 - acc: 0.6667
```

/04 [] ETA: 15c 0cc: 0.5055 0cc: 0.6449	17/04 [1 ETA: 14c Jacci 0 6215 acci 0 6672
	=	>
	-	=>] - ETA: 16s - loss: 0.6127 - acc: 0.6711
	=	=>] - ETA: 10s - loss: 0.6127 - acc: 0.6711
	-	=>] - ETA: 18s - loss: 0.6068 - acc: 0.6667
	-	=>] - ETA: 18s - loss: 0.5977 - acc: 0.6747
·		==>] - ETA: 195 - loss: 0.5963 - acc: 0.6753
	=	==>] - ETA: 173 loss: 0.5765 acc: 0.6745
	=	==>] - ETA: 173 loss: 0.5767 acc: 0.6745
	=	===>] - ETA: 20s - loss: 0.6004 - acc: 0.6779
	_	===>] - ETA: 203 - loss: 0.6056 - acc: 0.6748
		===>] - ETA: 20s - loss: 0.6052 - acc: 0.6730
	-	====>] - ETA: 20s - loss: 0.6063 - acc: 0.6724
/94 [====================================		====>] - ETA: 20s - loss: 0.6079 - acc: 0.6679
	=	====>
	=	====>] - ETA: 20s - loss: 0.5944 - acc: 0.6799
/94 [====================================	=	====>] - ETA: 20s - loss: 0.5744 - acc: 0.6777
	=	====>] - ETA: 20s - loss: 0.5730 - acc: 0.6782
	-	=====>] - ETA: 20s - loss: 0.5963 - acc: 0.6770
'94 [====================================		=====>] - ETA: 20s - loss: 0.5763 - acc: 0.6770
	=	=====>] - ETA: 20s - loss: 0.5970 - acc: 0.6767
/94 [====================================	-	=====>] - ETA: 20s - loss: 0.5772 - acc: 0.6767
/94 [====================================	-	======>] - ETA: 195 - loss: 0.5991 - acc: 0.6741
	=	======>] - ETA: 19s - loss: 0.5946 - acc: 0.6767
	=	======>] - ETA: 175 - loss: 0.5740 - acc: 0.6770
	-	======>] - ETA: 175 - loss: 0.5743 - acc: 0.6750
	_	======>] - ETA: 18s - loss: 0.5976 - acc: 0.6717
		======>] - ETA: 165 - 1055. 0.5770 - acc. 0.6717
·	=	======>] - ETA: 16s - loss: 0.6042 - acc: 0.6706
	-	-
•	=	=======>] - ETA: 18s - loss: 0.6021 - acc: 0.6693
•	=	=======>] - ETA: 17s - loss: 0.5978 - acc: 0.6704
	_	=======>
		=======>] - ETA: 17s - loss: 0.5956 - acc: 0.6736
		=======>] - ETA: 17s - loss: 0.5939 - acc: 0.6739
		=======>] - ETA: 16s - loss: 0.5941 - acc: 0.6735
	_	=======>] - ETA: 16s - loss: 0.5941 - acc: 0.6708
		=======>] - ETA: 16s - loss: 0.5917 - acc: 0.6711
	_	========>] - ETA: 15s - loss: 0.5942 - acc: 0.6691
94 [====================================		========>] - ETA: 15s - loss: 0.5958 - acc: 0.6689 =========>] - ETA: 15s - loss: 0.5969 - acc: 0.6698
	-	-
994	=	========>] - ETA: 14s - loss: 0.5965 - acc: 0.6695
	_	========>] - ETA: 14s - loss: 0.5936 - acc: 0.6725
		========>] - ETA: 14s - loss: 0.5925 - acc: 0.6739
94 [] - ETA: 15s - loss: 0.5968 - acc: 0.7188 94 [] - ETA: 15s - loss: 0.7291 - acc: 0.7031	-	======================================
94 [] - ETA: 15s - loss: 0.7291 - acc: 0.7031 94 [] - ETA: 14s - loss: 0.7061 - acc: 0.6875		
•	_	======================================
94 [>] - ETA: 13s - loss: 0.6867 - acc: 0.6484 94 [>] - ETA: 13s - loss: 0.6399 - acc: 0.6625		======================================
		======================================
		======================================
		======================================
		======================================
	-	======================================
	=	======================================
	=	======================================
	_	=======>] - ETA: 9s - loss: 0.5950 - acc: 0.6724
		=========>] - ETA: 9s - loss: 0.5950 - acc: 0.6722
	-	========>:] - ETA: 8s - loss: 0.5966 - acc: 0.6703
/94 [====>] - ETA: 13s - loss: 0.6220 - acc: 0.6641	/5/94 [====	==============>] - ETA: 8s - loss: 0.5983 - acc: 0.6718

```
38/94 [=======>:.....] - ETA: 20s - loss: 0.5769 - acc: 0.6871
39/94 [========>.....] - ETA: 20s - loss: 0.5744 - acc: 0.6863
40/94 [========>.....] - ETA: 19s - loss: 0.5733 - acc: 0.6879
41/94 [=======>.....] - ETA: 19s - loss: 0.5699 - acc: 0.6924
                                               42/94 [=======>:.....] - ETA: 19s - loss: 0.5677 - acc: 0.6960
43/94 [========>:....] - ETA: 19s - loss: 0.5691 - acc: 0.6951
44/94 [=======>.....] - ETA: 19s - loss: 0.5662 - acc: 0.6956
45/94 [========>:....] - ETA: 18s - loss: 0.5653 - acc: 0.6962
46/94 [=======>.....] - ETA: 18s - loss: 0.5698 - acc: 0.6946
47/94 [=======>.....] - ETA: 18s - loss: 0.5710 - acc: 0.6945
48/94 [=========>.....] - ETA: 17s - loss: 0.5737 - acc: 0.6937
49/94 [=========>.....] - ETA: 17s - loss: 0.5771 - acc: 0.6897
50/94 [=======>.....] - ETA: 17s - loss: 0.5744 - acc: 0.6915
52/94 [========>:.....] - ETA: 16s - loss: 0.5809 - acc: 0.6878
53/94 [=========>.....] - ETA: 16s - loss: 0.5834 - acc: 0.6866
54/94 [==========>.....] - ETA: 16s - loss: 0.5841 - acc: 0.6883
55/94 [=========>:....] - ETA: 15s - loss: 0.5851 - acc: 0.6844
56/94 [=========>......] - ETA: 15s - loss: 0.5840 - acc: 0.6861
57/94 [=========>.....] - ETA: 15s - loss: 0.5853 - acc: 0.6861
0.6585
                                               59/94 [=========>:.....] - ETA: 14s - loss: 0.5843 - acc: 0.6840
Epoch 18/20
                                               1/94 [......] - ETA: 16s - loss: 0.6439 - acc: 0.5625
2/94 [.....] - ETA: 15s - loss: 0.5770 - acc: 0.7031
                                               3/94 [......] - ETA: 14s - loss: 0.5729 - acc: 0.6875
                                               4/94 [>......] - ETA: 14s - loss: 0.5875 - acc: 0.6797
5/94 [>.....] - ETA: 13s - loss: 0.5640 - acc: 0.6750
                                               6/94 [>.....] - ETA: 12s - loss: 0.5771 - acc: 0.6587
7/94 [=>.....] - ETA: 12s - loss: 0.5764 - acc: 0.6628
                                               8/94 [=>.....] - ETA: 12s - loss: 0.5583 - acc: 0.6776
                                               9/94 [=>.....] - ETA: 11s - loss: 0.5605 - acc: 0.6856
                                               10/94 [==>.....] - ETA: 11s - loss: 0.5774 - acc: 0.6671
                                               11/94 [==>.....] - ETA: 11s - loss: 0.5912 - acc: 0.6576
                                               12/94 [==>.....] - ETA: 11s - loss: 0.5803 - acc: 0.6679
                                               13/94 [===>.....] - ETA: 11s - loss: 0.5794 - acc: 0.6766
                                               14/94 [===>.....] - ETA: 11s - loss: 0.5848 - acc: 0.6863
15/94 [===>.....] - ETA: 11s - loss: 0.5811 - acc: 0.6864
                                               16/94 [====>.....] - ETA: 13s - loss: 0.5789 - acc: 0.6864
                                               17/94 [====>.....] - ETA: 14s - loss: 0.5874 - acc: 0.6902
                                               18/94 [====>.....] - ETA: 15s - loss: 0.5845 - acc: 0.6900
                                               19/94 [====>.....] - ETA: 16s - loss: 0.5850 - acc: 0.6899
                                               20/94 [====>.....] - ETA: 17s - loss: 0.5811 - acc: 0.6898
                                               21/94 [====>.....] - ETA: 18s - loss: 0.5829 - acc: 0.6837
                                               22/94 [====>.....] - ETA: 18s - loss: 0.5880 - acc: 0.6754
                                               23/94 [=====>.....] - ETA: 19s - loss: 0.5832 - acc: 0.6786
                                               24/94 [=====>.....] - ETA: 19s - loss: 0.5789 - acc: 0.6829
25/94 [=====>.....] - ETA: 19s - loss: 0.5760 - acc: 0.6868
                                               26/94 [======>.....] - ETA: 20s - loss: 0.5714 - acc: 0.6857
                                               27/94 [======>.....] - ETA: 20s - loss: 0.5767 - acc: 0.6788
                                               28/94 [======>.....] - ETA: 20s - loss: 0.5800 - acc: 0.6824
                                               29/94 [======>.....] - ETA: 20s - loss: 0.5769 - acc: 0.6837
                                               30/94 [======>.....] - ETA: 20s - loss: 0.5761 - acc: 0.6828
                                               31/94 [======>.....] - ETA: 20s - loss: 0.5868 - acc: 0.6829
                                               32/94 [=======>.....] - ETA: 20s - loss: 0.5862 - acc: 0.6821
                                               33/94 [=======>.....] - ETA: 20s - loss: 0.5817 - acc: 0.6823
                                               93/94 [=======>:] - ETA: 0s - loss: 0.5874 - acc: 0.6789
34/94 [======>:....] - ETA: 20s - loss: 0.5823 - acc: 0.6824
35/94 [=======>.....] - ETA: 20s - loss: 0.5809 - acc: 0.6826
                                               36/94 [=======>.....] - ETA: 20s - loss: 0.5833 - acc: 0.6818
                                               37/94 [=======>.....] - ETA: 20s - loss: 0.5792 - acc: 0.6870
                                               0.6603
```

Epoch 19/20
1/94 [] - ETA: 16s - loss: 0.5151 - acc: 0.5625
2/94 [] - ETA: 15s - loss: 0.5331 - acc: 0.6406
3/94 [] - ETA: 14s - loss: 0.5549 - acc: 0.6562
4/94 [>] - ETA: 14s - loss: 0.5694 - acc: 0.6562
5/94 [>] - ETA: 13s - loss: 0.6004 - acc: 0.6562
6/94 [>] - ETA: 13s - loss: 0.6056 - acc: 0.6510
7/94 [=>] - ETA: 12s - loss: 0.6060 - acc: 0.6518
8/94 [=>] - ETA: 12s - loss: 0.5884 - acc: 0.6680
9/94 [=>] - ETA: 12s - loss: 0.5743 - acc: 0.6806
10/94 [==>] - ETA: 12s - loss: 0.5763 - acc: 0.6844
11/94 [==>] - ETA: 11s - loss: 0.5702 - acc: 0.6903
12/94 [==>] - ETA: 11s - loss: 0.5585 - acc: 0.6875
13/94 [===>] - ETA: 11s - loss: 0.5477 - acc: 0.6899
14/94 [===>] - ETA: 11s - loss: 0.5571 - acc: 0.6808
15/94 [===>] - ETA: 12s - loss: 0.5533 - acc: 0.6896
16/94 [====>] - ETA: 13s - loss: 0.5564 - acc: 0.6855
17/94 [====>] - ETA: 14s - loss: 0.5578 - acc: 0.6837
18/94 [====>] - ETA: 15s - loss: 0.5685 - acc: 0.6735
19/94 [=====>] - ETA: 16s - loss: 0.5722 - acc: 0.6759
20/94 [====>] - ETA: 17s - loss: 0.5666 - acc: 0.6733
21/94 [====>] - ETA: 17s - loss: 0.5686 - acc: 0.6695
22/94 [====>] - ETA: 18s - loss: 0.5702 - acc: 0.6689
23/94 [=====>] - ETA: 19s - loss: 0.5688 - acc: 0.6724
24/94 [=====>] - ETA: 19s - loss: 0.5697 - acc: 0.6757
25/94 [=====>] - ETA: 19s - loss: 0.5676 - acc: 0.6774
26/94 [======>] - ETA: 20s - loss: 0.5711 - acc: 0.6802
27/94 [======>] - ETA: 20s - loss: 0.5696 - acc: 0.6816
28/94 [======>] - ETA: 20s - loss: 0.5665 - acc: 0.6863
29/94 [======>] - ETA: 20s - loss: 0.5652 - acc: 0.6885
30/94 [=======>] - ETA: 20s - loss: 0.5631 - acc: 0.6895
31/94 [======>] - ETA: 20s - loss: 0.5657 - acc: 0.6884
32/94 [=======>] - ETA: 20s - loss: 0.5670 - acc: 0.6874
33/94 [=======>] - ETA: 20s - loss: 0.5605 - acc: 0.6903
34/94 [=======>] - ETA: 20s - loss: 0.5625 - acc: 0.6893
35/94 [=======>] - ETA: 20s - loss: 0.5605 - acc: 0.6901
36/94 [=======>] - ETA: 20s - loss: 0.5584 - acc: 0.6892
37/94 [=======>] - ETA: 20s - loss: 0.5576 - acc: 0.6891
38/94 [========>] - ETA: 20s - loss: 0.5596 - acc: 0.6891
39/94 [=======>] - ETA: 19s - loss: 0.5589 - acc: 0.6898
40/94 [=======>:] - ETA: 19s - loss: 0.5567 - acc: 0.6898
41/94 [=======>] - ETA: 19s - loss: 0.5548 - acc: 0.6935
42/94 [=======>>] - ETA: 19s - loss: 0.5580 - acc: 0.6927
43/94 [=======>>] - ETA: 19s - loss: 0.5604 - acc: 0.6940
44/94 [=======>>] - ETA: 18s - loss: 0.5616 - acc: 0.6938
45/94 [=======>] - ETA: 18s - loss: 0.5601 - acc: 0.6930
46/94 [======>] - ETA: 18s - loss: 0.5582 - acc: 0.6956
47/94 [====================================
48/94 [====================================
49/94 [====================================
50/94 [========================]
51/94 [====================================
52/94 [====================================
53/94 [=============]] - ETA: 16s - loss: 0.5592 - acc: 0.6945
54/94 [====================================
55/94 [====================================
50/94 [====================================
58/94 [====================================
50, / - [dect. 0.0720

```
59/94 [=========>:.....] - ETA: 14s - loss: 0.5577 - acc: 0.6917
64/94 [=========>......] - ETA: 12s - loss: 0.5587 - acc: 0.6880
0.6603
Epoch 19/20
1/94 [......] - ETA: 16s - loss: 0.5151 - acc: 0.5625
2/94 [......] - ETA: 15s - loss: 0.5331 - acc: 0.6406
3/94 [.....] - ETA: 14s - loss: 0.5549 - acc: 0.6562
4/94 [>.....] - ETA: 14s - loss: 0.5694 - acc: 0.6562
5/94 [>......] - ETA: 13s - loss: 0.6004 - acc: 0.6562
6/94 [>......] - ETA: 13s - loss: 0.6056 - acc: 0.6510
7/94 [=>.....] - ETA: 12s - loss: 0.6060 - acc: 0.6518
8/94 [=>.....] - ETA: 12s - loss: 0.5884 - acc: 0.6680
9/94 [=>.....] - ETA: 12s - loss: 0.5743 - acc: 0.6806
10/94 [==>.....] - ETA: 12s - loss: 0.5763 - acc: 0.6844
11/94 [==>.....] - ETA: 11s - loss: 0.5702 - acc: 0.6903
12/94 [==>.....] - ETA: 11s - loss: 0.5585 - acc: 0.6875
13/94 [===>.....] - ETA: 11s - loss: 0.5477 - acc: 0.6899
14/94 [===>.....] - ETA: 11s - loss: 0.5571 - acc: 0.6808
15/94 [===>.....] - ETA: 12s - loss: 0.5533 - acc: 0.6896
16/94 [====>.....] - ETA: 13s - loss: 0.5564 - acc: 0.6855
17/94 [====>.....] - ETA: 14s - loss: 0.5578 - acc: 0.6837
18/94 [====>.....] - ETA: 15s - loss: 0.5685 - acc: 0.6735
19/94 [====>.....] - ETA: 16s - loss: 0.5722 - acc: 0.6759
20/94 [====>.....] - ETA: 17s - loss: 0.5666 - acc: 0.6733
```

```
21/94 [====>.....] - ETA: 17s - loss: 0.5686 - acc: 0.6695
22/94 [====>.....] - ETA: 18s - loss: 0.5702 - acc: 0.6689
23/94 [=====>.....] - ETA: 19s - loss: 0.5688 - acc: 0.6724
24/94 [=====>.....] - ETA: 19s - loss: 0.5697 - acc: 0.6757
25/94 [=====>.....] - ETA: 19s - loss: 0.5676 - acc: 0.6774
26/94 [======>.....] - ETA: 20s - loss: 0.5711 - acc: 0.6802
27/94 [======>.....] - ETA: 20s - loss: 0.5696 - acc: 0.6816
28/94 [======>.....] - ETA: 20s - loss: 0.5665 - acc: 0.6863
29/94 [======>.....] - ETA: 20s - loss: 0.5652 - acc: 0.6885
30/94 [======>.....] - ETA: 20s - loss: 0.5631 - acc: 0.6895
31/94 [======>.....] - ETA: 20s - loss: 0.5657 - acc: 0.6884
32/94 [=======>.....] - ETA: 20s - loss: 0.5670 - acc: 0.6874
33/94 [======>.....] - ETA: 20s - loss: 0.5605 - acc: 0.6903
34/94 [======>.....] - ETA: 20s - loss: 0.5625 - acc: 0.6893
35/94 [=======>.....] - ETA: 20s - loss: 0.5605 - acc: 0.6901
36/94 [=======>.....] - ETA: 20s - loss: 0.5584 - acc: 0.6892
37/94 [=======>......] - ETA: 20s - loss: 0.5576 - acc: 0.6891
38/94 [=======>.....] - ETA: 20s - loss: 0.5596 - acc: 0.6891
39/94 [=======>.....] - ETA: 19s - loss: 0.5589 - acc: 0.6898
40/94 [========>.....] - ETA: 19s - loss: 0.5567 - acc: 0.6898
41/94 [=======>.....] - ETA: 19s - loss: 0.5548 - acc: 0.6935
42/94 [=======>.....] - ETA: 19s - loss: 0.5580 - acc: 0.6927
43/94 [=======>.....] - ETA: 19s - loss: 0.5604 - acc: 0.6940
44/94 [========>.....] - ETA: 18s - loss: 0.5616 - acc: 0.6938
45/94 [=======>.....] - ETA: 18s - loss: 0.5601 - acc: 0.6930
46/94 [=======>.....] - ETA: 18s - loss: 0.5582 - acc: 0.6956
47/94 [=======>:....] - ETA: 18s - loss: 0.5613 - acc: 0.6948
48/94 [=======>.....] - ETA: 17s - loss: 0.5596 - acc: 0.6966
49/94 [=========>.....] - ETA: 17s - loss: 0.5589 - acc: 0.6951
50/94 [=======>.....] - ETA: 17s - loss: 0.5599 - acc: 0.6925
51/94 [========>.....] - ETA: 16s - loss: 0.5611 - acc: 0.6917
52/94 [========>......] - ETA: 16s - loss: 0.5599 - acc: 0.6923
53/94 [========>:....] - ETA: 16s - loss: 0.5592 - acc: 0.6945
56/94 [=========>.....] - ETA: 15s - loss: 0.5572 - acc: 0.6925
57/94 [=========>......] - ETA: 14s - loss: 0.5582 - acc: 0.6929
58/94 [=========>............] - ETA: 14s - loss: 0.5575 - acc: 0.6928
59/94 [=========>......] - ETA: 14s - loss: 0.5577 - acc: 0.6917
60/94 [==========>......] - ETA: 13s - loss: 0.5573 - acc: 0.6906
67/94 [==========>........] - ETA: 11s - loss: 0.5569 - acc: 0.6879
69/94 [==========>.......] - ETA: 10s - loss: 0.5560 - acc: 0.6870
```

```
92/94 [======>:] - ETA: 1s - loss: 0.5607 - acc: 0.6858
0.6607
Epoch 20/20
1/94 [......] - ETA: 16s - loss: 0.7178 - acc: 0.5625
2/94 [......] - ETA: 15s - loss: 0.6425 - acc: 0.6250
4/94 [>.....] - ETA: 13s - loss: 0.5649 - acc: 0.6953
5/94 [>.....] - ETA: 13s - loss: 0.5892 - acc: 0.6625
6/94 [>.....] - ETA: 13s - loss: 0.5499 - acc: 0.6979
7/94 [=>.....] - ETA: 12s - loss: 0.5410 - acc: 0.6875
8/94 [=>.....] - ETA: 12s - loss: 0.5341 - acc: 0.6953
9/94 [=>.....] - ETA: 12s - loss: 0.5382 - acc: 0.6875
10/94 [==>.....] - ETA: 11s - loss: 0.5272 - acc: 0.6969
11/94 [==>.....] - ETA: 11s - loss: 0.5180 - acc: 0.7017
12/94 [==>.....] - ETA: 11s - loss: 0.5188 - acc: 0.7005
13/94 [===>.....] - ETA: 11s - loss: 0.5159 - acc: 0.7043
14/94 [===>.....] - ETA: 11s - loss: 0.5153 - acc: 0.7054
15/94 [===>.....] - ETA: 12s - loss: 0.5248 - acc: 0.7000
16/94 [====>.....] - ETA: 13s - loss: 0.5330 - acc: 0.6992
17/94 [====>.....] - ETA: 14s - loss: 0.5397 - acc: 0.6967
18/94 [====>.....] - ETA: 15s - loss: 0.5417 - acc: 0.6979
19/94 [====>.....] - ETA: 16s - loss: 0.5453 - acc: 0.6875
20/94 [====>.....] - ETA: 17s - loss: 0.5471 - acc: 0.6859
21/94 [====>.....] - ETA: 18s - loss: 0.5429 - acc: 0.6845
22/94 [====>.....] - ETA: 18s - loss: 0.5350 - acc: 0.6889
23/94 [=====>.....] - ETA: 19s - loss: 0.5345 - acc: 0.6889
24/94 [=====>.....] - ETA: 19s - loss: 0.5396 - acc: 0.6888
25/94 [=====>.....] - ETA: 19s - loss: 0.5384 - acc: 0.6925
26/94 [======>.....] - ETA: 20s - loss: 0.5372 - acc: 0.6935
27/94 [======>.....] - ETA: 20s - loss: 0.5325 - acc: 0.6991
28/94 [======>.....] - ETA: 20s - loss: 0.5269 - acc: 0.7042
29/94 [======>.....] - ETA: 20s - loss: 0.5231 - acc: 0.7047
30/94 [======>......] - ETA: 20s - loss: 0.5314 - acc: 0.7063
31/94 [======>.....] - ETA: 20s - loss: 0.5329 - acc: 0.7046
32/94 [=======>......] - ETA: 20s - loss: 0.5332 - acc: 0.7012
33/94 [======>:.....] - ETA: 20s - loss: 0.5321 - acc: 0.7008
34/94 [======>:....] - ETA: 20s - loss: 0.5283 - acc: 0.7013
35/94 [=======>.....] - ETA: 20s - loss: 0.5254 - acc: 0.7018
36/94 [=======>......] - ETA: 20s - loss: 0.5219 - acc: 0.7040
37/94 [=======>.....] - ETA: 20s - loss: 0.5225 - acc: 0.7052
38/94 [=======>:.....] - ETA: 19s - loss: 0.5245 - acc: 0.7031
39/94 [=======>.....] - ETA: 19s - loss: 0.5278 - acc: 0.7003
40/94 [=======>.....] - ETA: 19s - loss: 0.5321 - acc: 0.6977
41/94 [=======>.....] - ETA: 19s - loss: 0.5315 - acc: 0.7005
```

```
42/94 [========>.....] - ETA: 19s - loss: 0.5315 - acc: 0.7024
43/94 [=========>......] - ETA: 18s - loss: 0.5317 - acc: 0.7028
44/94 [========>......] - ETA: 18s - loss: 0.5338 - acc: 0.7003
45/94 [=======>.....] - ETA: 18s - loss: 0.5349 - acc: 0.6993
46/94 [=======>.....] - ETA: 18s - loss: 0.5364 - acc: 0.6990
47/94 [========>.....] - ETA: 18s - loss: 0.5397 - acc: 0.6968
48/94 [========>.....] - ETA: 17s - loss: 0.5406 - acc: 0.6947
49/94 [=======>:....] - ETA: 17s - loss: 0.5384 - acc: 0.6977
50/94 [========>.....] - ETA: 17s - loss: 0.5392 - acc: 0.6953
51/94 [========>.....] - ETA: 16s - loss: 0.5401 - acc: 0.6970
52/94 [========>:....] - ETA: 16s - loss: 0.5385 - acc: 0.6980
53/94 [=========>.....] - ETA: 16s - loss: 0.5365 - acc: 0.6990
55/94 [=========>......] - ETA: 15s - loss: 0.5408 - acc: 0.6946
56/94 [=========>.....] - ETA: 15s - loss: 0.5394 - acc: 0.6956
57/94 [=========>......] - ETA: 14s - loss: 0.5377 - acc: 0.6965
58/94 [=========>......] - ETA: 14s - loss: 0.5379 - acc: 0.6958
59/94 [========>:.....] - ETA: 14s - loss: 0.5377 - acc: 0.6952
60/94 [==========>......] - ETA: 13s - loss: 0.5397 - acc: 0.6961
61/94 [==========>:.....] - ETA: 13s - loss: 0.5385 - acc: 0.6970
65/94 [=========>......] - ETA: 12s - loss: 0.5366 - acc: 0.6945
67/94 [=========>:......] - ETA: 11s - loss: 0.5400 - acc: 0.6956
69/94 [=========>:.......] - ETA: 10s - loss: 0.5428 - acc: 0.6940
```

```
95/94 [========================] - 186s 2s/step - loss: 0.5430 - acc: 0.6952 - val_loss: 0.6259 - val_acc:
0.6426
Test loss: 6.40280550737
Test accuracy 0.587692307876
Module completed; cleaning up.
Clean up finished.
Task completed successfully.
I Tearing down TensorFlow.
I Finished tearing down TensorFlow.
03:33:16.526 Job completed successfully
```

```
Cloud ML Job, cnncopy_train_20171109_02... 
All logs
                                       ▼ Any log level ▼ Jump to date ▼
                                                                                                             View Options
▶ 1 03:2/:12.398 master-reptica-0 85/94 [===============....] - EIA: 4s - Loss: 0.5455 - acc: 0.6924
:
▶ 1 03:27:13.368 master-replica-0 87/94 [========>...] - ETA: 3s - loss: 0.5432 - acc: 0.6941
▶ 1 03:27:13.851 master-replica-0 88/94 [==================...] - ETA: 2s - loss: 0.5435 - acc: 0.6930
> 3:27:15.796 master-replica-0 92/94 [====================], - ETA: 1s - loss: 0.5432 - acc: 0.6941
▶ 1 03:27:16.285 master-replica-0 93/94 [==============>,] - ETA: 0s - loss: 0.5423 - acc: 0.6944
▶ 1 03:29:42.171 master-replica-0 95/94 [=========] - 1865 2s/step - loss: 0.5430 - acc: 0.6952 - val loss: 0.6259 - val acc: 0.6426
▶ 🔼 03:29:47.980 master-replica-0 Test loss: 6.40280550737
▶ 1 03:29:47.981 master-replica-0 Test accuracy 0.587692307876
▶ 3 03:29:48.076 master-replica-0 Module completed; cleaning up.
▶ 1 03:29:48.076 master-replica-0 Clean up finished.
▶ ■ 03:29:48.077 master-replica-0 Task completed successfully.
▶ 3 03:30:04.818 Tearing down TensorFlow.
▶ 1 03:32:02.147 Finished tearing down TensorFlow
▶ ■ 03:33:16.526 Job completed successfully
11/9
1/20
       loss: 1.3712 - acc: 0.2991 - val loss: 1.3440 - val acc: 0.3217 master-replica-0
      loss: 1.3037 - acc: 0.3508 - val loss: 1.2360 - val acc: 0.3800 master-replica-0
2/20
3/20
      loss: 1.2214 - acc: 0.3915 - val loss: 1.1272 - val acc: 0.4431 master-replica-0
       loss: 1.0646 - acc: 0.4565 - val loss: 0.9222 - val acc: 0.5043 master-replica-0
4/20
5/20
```

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loss: 0.9621 - acc: 0.5140 - val loss: 0.9246 - val acc: 0.5083 master-replica-0
       loss: 0.8789 - acc: 0.5333 - val loss: 0.7766 - val acc: 0.5766 master-replica-0
6/20
7/20
        loss: 0.8278 - acc: 0.5612 - val loss: 0.8265 - val acc: 0.5536 master-replica-0
8/20
        loss: 0.7757 - acc: 0.5858 - val loss: 0.7457 - val acc: 0.5831 master-replica-0
9/20
       loss: 0.7514 - acc: 0.5916 - val loss: 0.7268 - val acc: 0.6072 master-replica-0
        loss: 0.7049 - acc: 0.6246 - val loss: 0.6672 - val acc: 0.6261 master-replica-0
10/20
        loss: 0.6871 - acc: 0.6238 - val loss: 0.6472 - val acc: 0.6340 master-replica-0
11/20
12/20
       loss: 0.6725 - acc: 0.6426 - val loss: 0.6551 - val acc: 0.6422 master-replica-0
13/20
        loss: 0.6370 - acc: 0.6405 - val loss: 0.5923 - val acc: 0.6561 master-replica-0
14/20
        loss: 0.6250 - acc: 0.6527 - val loss: 0.6110 - val acc: 0.6447 master-replica-0
15/20
        loss: 0.6126 - acc: 0.6515 - val loss: 0.6138 - val acc: 0.6541 master-replica-0
       loss: 0.5811 - acc: 0.6746 - val loss: 0.8016 - val acc: 0.5994 master-replica-0
16/20
17/20
        loss: 0.5870 - acc: 0.6767 - val loss: 0.5910 - val acc: 0.6585 master-replica-0
       loss: 0.5855 - acc: 0.6801 - val loss: 0.5970 - val acc: 0.6603 master-replica-0
18/20
19/20
       loss: 0.5637 - acc: 0.6829 - val loss: 0.6003 - val acc: 0.6607 master-replica-0
       loss: 0.5430 - acc: 0.6952 - val loss: 0.6259 - val acc: 0.6426 master-replica-0
Test loss: 6.40280550737 master-replica-0
Test accuracy 0.587692307876 master-replica-0
```