```
In [5]: import pandas as pd
         import numpy as np
         import tensorflow as tf
         import matplotlib.pyplot as plt
          from tensorflow import keras
         import pickle
          from dataloader import tfdata_generator
          from keras.models import load_model
         import os
 In [4]: labels = pd.read_csv('labels.csv')
         labels.head()
 Out[4]:
                                              breed
          0 000bec180eb18c7604dcecc8fe0dba07
                                           boston_bull
              001513dfcb2ffafc82cccf4d8bbaba97
                                               dingo
          2 001cdf01b096e06d78e9e5112d419397
                                             pekinese
          3 00214f311d5d2247d5dfe4fe24b2303d
                                              bluetick
              0021f9ceb3235effd7fcde7f7538ed62 golden_retriever
 In [6]: files_path = []
          for root, dirs, files in os.walk("train", topdown=False):
             for name in sorted(files):
                files_path.append(os.path.join(root, name))
 In [7]: files = pd.DataFrame(list(zip(files_path, labels.breed.to_list())), columns=['path', 'label'])
          files.head()
 Out[7]:
                                         path
                                                     label
          train/000bec180eb18c7604dcecc8fe0dba07.jpg
                                                 boston_bull
              train/001513dfcb2ffafc82cccf4d8bbaba97.jpg
                                                     dingo
          2 train/001cdf01b096e06d78e9e5112d419397.jpg
                                                   pekinese
             train/00214f311d5d2247d5dfe4fe24b2303d.jpg
              train/0021f9ceb3235effd7fcde7f7538ed62.jpg golden_retriever
 In [8]: #Loading the encoder which is created while training.
          with open('/content/drive/MyDrive/dog_breed/encoder','rb') as file:
              encoder = pickle.load(file)
 In [9]: image_labels = encoder.transform(files['label'].to_list())
In [10]: image_paths = files['path'].tolist()
In [11]: image_data_generator = tfdata_generator(image_paths, image_labels, is_training=False)
In [12]: #Loading the best model obtained after experimentation
          Stacked_model_1 = load_model('/content/drive/MyDrive/dog_breed/Inception_resnet_stacked_weights.h5')
In [13]: | def modelling(data_generator, model=Stacked_model_1):
              This function will return the Loss and Accuracy by evaluating the data using the best model.
              This function takes train data as tensor and trained model.
                              = model.evaluate(data_generator, steps=100, verbose=0)
              results
              print('Loss: ',results[0])
              print('Accuracy: ',results[1])
              return None
          modelling(image_data_generator)
         Loss: 0.1003708466887474
         Accuracy: 0.9706249833106995
In [14]: test_files = []
          for root, dirs, files in os.walk("test", topdown=False):
             for name in sorted(files):
                test_files.append(os.path.join(root, name))
In [15]: test_image_path = test_files[0]
In [16]: def estimate(test_image_path,model=Stacked_model_1):
              This function will return the predictions for each dog breed label
              by taking the image and best model as the input.
              image = tf.io.read_file(test_image_path)
              image = tf.image.decode_jpeg(image, channels=3)
              image = image/255
              image = tf.image.convert_image_dtype(image, tf.float32)
              image = tf.image.resize(image, [400, 400])
              predictions = model.predict(image[np.newaxis,:,:,:],verbose=0)
              return predictions
          estimate(test_image_path)
Out[16]: array([[2.13981949e-07, 1.09429372e-07, 8.95456296e-08, 1.00257083e-07,
                  3.71944026e-07, 1.68061689e-08, 2.82568102e-08, 3.74299098e-08,
                  1.00755123e-07, 4.33248175e-08, 2.38967264e-08, 5.71410013e-08,
                  8.49853876e-08, 2.27446276e-06, 7.34796288e-07, 5.02358830e-07,
                  2.92693187e-07, 2.56265924e-08, 7.31431626e-08, 6.30322745e-07,
                  1.25154926e-08, 2.10035353e-07, 1.56985288e-06, 2.64991769e-07,
                  4.32793925e-08, 2.14196052e-06, 3.56475567e-07, 4.01320079e-08,
                  2.72786508e-07, 2.94776669e-06, 3.16347666e-07, 6.17105167e-09,
                  3.75244980e-07, 3.38983938e-07, 3.27307902e-07, 1.49979962e-06,
                  1.39323265e-06, 4.38105019e-09, 6.78227536e-07, 4.28247882e-09,
                  2.79466462e-07, 3.17898703e-08, 6.08308071e-08, 9.54548707e-09,
                  1.33346916e-07, 1.96643285e-07, 1.04513475e-07, 2.16150070e-07,
                  6.23792857e-08, 1.04177877e-06, 1.56781439e-07, 4.94047697e-07,
                  8.49129620e-08, 1.32094158e-08, 3.57634491e-08, 1.81291853e-08,
                  4.89738841e-06, 7.56207328e-08, 2.34532589e-08, 2.63651838e-08,
                  2.24527099e-08, 9.98915315e-01, 1.66896381e-08, 8.32304110e-08,
                  7.12395121e-08, 2.27548540e-07, 2.51263884e-07, 1.44957923e-07,
                  1.35459985e-07, 3.85751449e-08, 3.99500323e-07, 4.42862330e-07,
                  1.67443133e-07, 2.71461568e-06, 6.41320810e-07, 2.04694985e-07,
                  2.18815629e-07, 5.20136609e-07, 4.40689867e-08, 3.47516966e-08,
                  1.44090009e-07, 4.27515161e-08, 1.22546425e-07, 3.22490621e-08,
                  2.13560452e-06, 1.00575329e-03, 2.87706456e-07, 1.56010174e-07,
                  7.37635617e-07, 1.68240746e-07, 1.62158176e-06, 6.43870450e-08,
                  2.82288397e-07, 3.23727363e-07, 1.41815121e-07, 1.99424193e-07,
                  8.77574067e-08, 3.77830283e-08, 4.20133404e-07, 1.58100408e-06,
                  3.00330012e-05, 5.75637955e-08, 1.88120026e-08, 1.90882247e-07,
                  1.45807547e-07, 2.89138427e-07, 1.82370030e-07, 3.24190310e-08,
                  1.76751499e-07, 3.89571960e-06, 1.06400755e-07, 4.01414411e-07,
                  1.62019358e-08, 9.64391091e-08, 1.01823218e-06, 7.00956022e-08,
                  3.60562353e-07, 1.14011813e-08, 9.01388830e-08, 2.44355931e-07]],
                dtype=float32)
```