```
import itertools
import os

import keras
from keras.applications.inception_v3 import Inceptionv3
from keras.applications.inception_v3 import preprocess_input as
inception_v3_preprocessor
from keras.layers import Dense, GlobalAveragePooling2D, Input
from keras.preprocessing import image
from keras.preprocessing.image import ImageDataGenerator
from keras.utils import np_utils
import matplotlib.pyplot as plt
import numpy as np
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import train_test_split
from tqdm import tqdm
```

Using TensorFlow backend.

%matplotlib inline

```
INPUT_SIZE = 299
```

```
data_dir = '../data/crop/'
label_path = '../breeds_16.txt'

breeds = []
with open(label_path, 'r') as f:
    lines = f.readlines()
    for line in lines:
        breeds.append(line.replace('\n', ''))
print(len(breeds))
```

```
def load_img(img_path):
    """ Load a image with (INPUT_SIZE, INPUT_SIZE)
    Returns:
        RGB image as numpy array
    """
    img = image.load_img(img_path, target_size=(INPUT_SIZE, INPUT_SIZE))
    img = image.img_to_array(img)
    img /= 255
    return img
```

```
# Load all images and labels
X = []
y = []
for breed in breeds:
    print(breed)
    index = breeds.index(breed)
    imgs = os.listdir(data_dir + breed + '/')
    for img in imgs:
        img_path = data_dir + breed + '/' + img
        X.append(load_img(img_path))
        y.append(index)
```

```
beagle
border_collie
borzoi
chihuahua
doberman
french_bulldog
german_shepherd
golden_retriever
maltese_dog
pembroke
pomeranian
pug
shih-tzu
siberian_husky
toy_poodle
yorkshire_terrier
```

```
X = np.array(X)
y = np.array(y)
```

```
y = np_utils.to_categorical(y)
print(X.shape, y.shape)
```

```
(1360, 299, 299, 3) (1360, 16)
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
print(X_train.shape, y_train.shape, X_test.shape, y_test.shape)
(952, 299, 299, 3) (952, 16) (408, 299, 299, 3) (408, 16)
# Data augmentation
train_datagen = ImageDataGenerator(rotation_range=30,
                                   zoom\_range = 0.1,
                                   width_shift_range=0.2,
                                   height_shift_range=0.2,
                                   horizontal_flip = True)
train_generator = train_datagen.flow(X_train, y_train, shuffle=False, batch_size=10,
seed=10)
base_model = Inceptionv3(weights='imagenet', include_top=False, input_shape=(299, 299,
3))
""" Transfer learning """
x = base\_model.output
x = GlobalAveragePooling2D()(x)
x = Dense(512, activation='relu')(x)
prediction = Dense(len(breeds), activation='softmax')(x)
model = keras.models.Model(inputs=base_model.input, outputs=prediction)
for layer in base_model.layers:
    layer.trainable=False
model.compile(optimizer=keras.optimizers.Adam(lr=0.0001),
              loss='categorical_crossentropy',
              metrics=['accuracy'])
model.summary()
Layer (type)
                                Output Shape
                                                   Param #
                                                                 Connected to
input_1 (InputLayer) (None, 299, 299, 3) 0
```

conv2d_1 (Conv2D)	(None,	149, 14	19, 32	) 864	input_1[0][0]
 batch_normalization_1 (BatchNor	(None,	149, 14	19, 32	) 96	conv2d_1[0][0]
activation_1 (Activation) batch_normalization_1[0][0]	(None,	149, 14	19, 32	) 0	
conv2d_2 (Conv2D)	(None,	147, 14	17, 32	) 9216	activation_1[0][0]
batch_normalization_2 (BatchNor	(None,	147, 14	17, 32	) 96	conv2d_2[0][0]
activation_2 (Activation) batch_normalization_2[0][0]	(None,	147, 14	17, 32	) 0	
conv2d_3 (Conv2D)	(None,	147, 14	17, 64	) 18432	activation_2[0][0]
batch_normalization_3 (BatchNor	(None,	147, 14	17, 64	) 192	conv2d_3[0][0]
activation_3 (Activation) batch_normalization_3[0][0]	(None,	147, 14	17, 64	) 0	
max_pooling2d_1 (MaxPooling2D)	(None,	73, 73,	, 64)	0	activation_3[0][0]
conv2d_4 (Conv2D)	(None,	73, 73,	, 80)	5120	max_pooling2d_1[0][0]
  batch_normalization_4 (BatchNor	(None,	73, 73,	, 80)	240	conv2d_4[0][0]
activation_4 (Activation) batch_normalization_4[0][0]	(None,	73, 73,	, 80)	0	

conv2d_5 (Conv2D)	(None,	71,	71,	192)	138240	activation_4[0][0]
batch_normalization_5 (BatchNor	(None,	71,	71,	192)	576	conv2d_5[0][0]
activation_5 (Activation) batch_normalization_5[0][0]	(None,	71,	71,	192)	0	
max_pooling2d_2 (MaxPooling2D)	(None,	35,	35,	192)	0	activation_5[0][0]
conv2d_9 (Conv2D)	(None,	35,	35,	64)	12288	max_pooling2d_2[0][0]
batch_normalization_9 (BatchNor	(None,	35,	35,	64)	192	conv2d_9[0][0]
activation_9 (Activation) batch_normalization_9[0][0]	(None,	35,	35,	64)	0	
conv2d_7 (Conv2D)	(None,	35,	35,	48)	9216	max_pooling2d_2[0][0]
conv2d_10 (Conv2D)	(None,	35,	35,	96)	55296	activation_9[0][0]
batch_normalization_7 (BatchNor	(None,	35,	35,	48)	144	conv2d_7[0][0]
batch_normalization_10 (BatchNo	(None,	35,	35,	96)	288	conv2d_10[0][0]
activation_7 (Activation) batch_normalization_7[0][0]	(None,	35,	35,	48)	0	
activation_10 (Activation) batch_normalization_10[0][0]	(None,	35,	35,	96)	0	

agePoo (Non	e, 35,	35,	192)	0	max_pooling2d_2[0][0]
(Non	e, 35,	35,	64)	12288	max_pooling2d_2[0][0]
(Non	e, 35,	35,	64)	76800	activation_7[0][0]
(Non	e, 35,	35,	96)	82944	activation_10[0][0]
(Non	e, 35,	35,	32)	6144	average_pooling2d_1[0]
tchNor (Non	e, 35,	35,	64)	192	conv2d_6[0][0]
tchNor (Non	e, 35,	35,	64)	192	conv2d_8[0][0]
atchNo (Non	e, 35,	35,	96)	288	conv2d_11[0][0]
atchNo (Non	e, 35,	35,	32)	96	conv2d_12[0][0]
	e, 35,	35,	64)	0	
	e, 35,	35,	64)	0	
	e, 35,	35,	96)	0	
	e, 35,	35,	32)	0	
3 3 3	(None (None (None atchnor (None Batchno (None (N	(None, 35,  (None, 35,  (None, 35,  (None, 35,  atchNor (None, 35,  BatchNo (None, 35,  (N	(None, 35, 35,  (None, 35, 35,  (None, 35, 35,  (None, 35, 35,  atchnor (None, 35, 35,  atchno (None, 35, 35,  (None, 35, 35,	(None, 35, 35, 64)  (None, 35, 35, 64)  (None, 35, 35, 96)  (None, 35, 35, 64)  atchnor (None, 35, 35, 64)  Batchno (None, 35, 35, 64)  (None, 35, 35, 64)	(None, 35, 35, 96) 82944  (None, 35, 35, 32) 6144  (None, 35, 35, 64) 192  (RatchNor (None, 35, 35, 64) 192  (RatchNo (None, 35, 35, 96) 288  (None, 35, 35, 64) 0  (None, 35, 35, 64) 0

mixedO (Concatenate)	(None,	35,	35,	256)	0	activation_6[0][0] activation_8[0][0]
						<pre>activation_11[0][0] activation_12[0][0]</pre>
conv2d_16 (Conv2D)	(None,	35,	35,	64)	16384	mixed0[0][0]
batch_normalization_16 (BatchNo	(None,	35,	35,	64)	192	conv2d_16[0][0]
activation_16 (Activation) batch_normalization_16[0][0]	(None,	35,	35,	64)	0	
conv2d_14 (Conv2D)	(None,	35,	35,	48)	12288	mixed0[0][0]
conv2d_17 (Conv2D)	(None,	35,	35,	96)	55296	activation_16[0][0]
batch_normalization_14 (BatchNo	(None,	35,	35,	48)	144	conv2d_14[0][0]
batch_normalization_17 (BatchNo	(None,	35,	35,	96)	288	conv2d_17[0][0]
activation_14 (Activation) batch_normalization_14[0][0]	(None,	35,	35,	48)	0	
activation_17 (Activation) batch_normalization_17[0][0]	(None,	35,	35,	96)	0	
average_pooling2d_2 (AveragePoo	(None,	35,	35,	256)	0	mixed0[0][0]

conv2d_13 (Conv2D)	(None,	35,	35,	64)	16384	mixed0[0][0]
conv2d_15 (Conv2D)	(None,	35,	35,	64)	76800	activation_14[0][0]
conv2d_18 (Conv2D)	(None,	35,	35,	96)	82944	activation_17[0][0]
 conv2d_19 (Conv2D) [0]	(None,	35,	35,	64)	16384	average_pooling2d_2[0]
batch_normalization_13 (BatchNo	(None,	35,	35,	64)	192	conv2d_13[0][0]
batch_normalization_15 (BatchNo	(None,	35,	35,	64)	192	conv2d_15[0][0]
  batch_normalization_18 (BatchNo	(None,	35,	35,	96)	288	conv2d_18[0][0]
  batch_normalization_19 (BatchNo	(None,	35,	35,	64)	192	conv2d_19[0][0]
activation_13 (Activation) batch_normalization_13[0][0]	(None,	35,	35,	64)	0	
activation_15 (Activation) batch_normalization_15[0][0]	(None,	35,	35,	64)	0	
activation_18 (Activation) batch_normalization_18[0][0]	(None,	35,	35,	96)	0	
activation_19 (Activation) batch_normalization_19[0][0]	(None,	35,	35,	64)	0	
mixed1 (Concatenate)	(None,	35,	35,	288)	0	activation_13[0][0] activation_15[0][0]

						activation_18[0][0] activation_19[0][0]
conv2d_23 (Conv2D)	(None,	35,	35,	64)	18432	mixed1[0][0]
batch_normalization_23 (BatchNo	(None,	35,	35,	64)	192	conv2d_23[0][0]
activation_23 (Activation) batch_normalization_23[0][0]	(None,	35,	35,	64)	0	
conv2d_21 (Conv2D)	(None,	35,	35,	48)	13824	mixed1[0][0]
conv2d_24 (Conv2D)	(None,	35,	35,	96)	55296	activation_23[0][0]
batch_normalization_21 (BatchNo	(None,	35,	35,	48)	144	conv2d_21[0][0]
batch_normalization_24 (BatchNo	(None,	35,	35,	96)	288	conv2d_24[0][0]
activation_21 (Activation) batch_normalization_21[0][0]	(None,	35,	35,	48)	0	
activation_24 (Activation) batch_normalization_24[0][0]	(None,	35,	35,	96)	0	
average_pooling2d_3 (AveragePoo	(None,	35,	35,	288)	0	mixed1[0][0]
conv2d_20 (Conv2D)	(None,	35,	35,	64)	18432	mixed1[0][0]
conv2d_22 (Conv2D)	(None,	35,	35,	64)	76800	activation_21[0][0]

conv2d_25 (Conv2D)	(None,	35,	35,	96)	82944	activation_24[0][0]
conv2d_26 (Conv2D) [0]	(None,	35,	35,	64)	18432	average_pooling2d_3[0]
batch_normalization_20 (BatchNo	(None,	35,	35,	64)	192	conv2d_20[0][0]
batch_normalization_22 (BatchNo	(None,	35,	35,	64)	192	conv2d_22[0][0]
  batch_normalization_25 (BatchNo	(None,	35,	35,	96)	288	conv2d_25[0][0]
  batch_normalization_26 (BatchNo	(None,	35,	35,	64)	192	conv2d_26[0][0]
activation_20 (Activation) batch_normalization_20[0][0]	(None,	35,	35,	64)	0	
activation_22 (Activation) batch_normalization_22[0][0]	(None,	35,	35,	64)	0	
activation_25 (Activation) batch_normalization_25[0][0]	(None,	35,	35,	96)	0	
activation_26 (Activation) batch_normalization_26[0][0]	(None,	35,	35,	64)	0	
mixed2 (Concatenate)	(None,	35,	35,	288)	0	activation_20[0][0]
						activation_22[0][0]
						activation_25[0][0]
						activation_26[0][0]

conv2d_28 (Conv2D)	(None,	35,	35,	64)	18432	mixed2[0][0]
batch_normalization_28 (BatchNo	(None,	35,	35,	64)	192	conv2d_28[0][0]
activation_28 (Activation) batch_normalization_28[0][0]	(None,	35,	35,	64)	0	
conv2d_29 (Conv2D)	(None,	35,	35,	96)	55296	activation_28[0][0]
batch_normalization_29 (BatchNo	(None,	35,	35,	96)	288	conv2d_29[0][0]
activation_29 (Activation) batch_normalization_29[0][0]	(None,	35,	35,	96)	0	
conv2d_27(Conv2D)	(None,	17,	17,	384)	995328	mixed2[0][0]
conv2d_30 (Conv2D)	(None,	17,	17,	96)	82944	activation_29[0][0]
batch_normalization_27 (BatchNo	(None,	17,	17,	384)	1152	conv2d_27[0][0]
batch_normalization_30 (BatchNo	(None,	17,	17,	96)	288	conv2d_30[0][0]
activation_27 (Activation) batch_normalization_27[0][0]	(None,	17,	17,	384)	0	
activation_30 (Activation) batch_normalization_30[0][0]	(None,	17,	17,	96)	0	
max_pooling2d_3 (MaxPooling2D)	(None,	17,	17,	288)	0	mixed2[0][0]

mixed3 (Concatenate)	(None,	17,	17,	768)	0	<pre>activation_27[0][0] activation_30[0][0] max_pooling2d_3[0][0]</pre>
conv2d_35 (Conv2D)	(None,	17,	17,	128)	98304	mixed3[0][0]
batch_normalization_35 (BatchNo	(None,	17,	17,	128)	384	conv2d_35[0][0]
activation_35 (Activation) batch_normalization_35[0][0]	(None,	17,	17,	128)	0	
conv2d_36 (Conv2D)	(None,	17,	17,	128)	114688	activation_35[0][0]
batch_normalization_36 (BatchNo	(None,	17,	17,	128)	384	conv2d_36[0][0]
activation_36 (Activation) batch_normalization_36[0][0]	(None,	17,	17,	128)	0	
conv2d_32 (Conv2D)	(None,	17,	17,	128)	98304	mixed3[0][0]
conv2d_37 (Conv2D)	(None,	17,	17,	128)	114688	activation_36[0][0]
batch_normalization_32 (BatchNo	(None,	17,	17,	128)	384	conv2d_32[0][0]
batch_normalization_37 (BatchNo	(None,	17,	17,	128)	384	conv2d_37[0][0]
activation_32 (Activation) batch_normalization_32[0][0]	(None,	17,	17,	128)	0	

activation_37 (Activation) batch_normalization_37[0][0]	(None,	17,	17,	128)	0	
conv2d_33 (Conv2D)	(None,	17,	17,	128)	114688	activation_32[0][0]
 conv2d_38 (Conv2D)	(None,	17,	17,	128)	114688	activation_37[0][0]
  batch_normalization_33 (BatchNo	(None,	17,	17,	128)	384	conv2d_33[0][0]
  batch_normalization_38 (BatchNo	(None,	17,	17,	128)	384	conv2d_38[0][0]
activation_33 (Activation) batch_normalization_33[0][0]	(None,	17,	17,	128)	0	
activation_38 (Activation) batch_normalization_38[0][0]	(None,	17,	17,	128)	0	
average_pooling2d_4 (AveragePoo	(None,	17,	17,	768)	0	mixed3[0][0]
conv2d_31 (Conv2D)	(None,	17,	17,	192)	147456	mixed3[0][0]
conv2d_34 (Conv2D)	(None,	17,	17,	192)	172032	activation_33[0][0]
  conv2d_39 (Conv2D)	(None,	17,	17,	192)	172032	activation_38[0][0]
 conv2d_40 (Conv2D) [0]	(None,	17,	17,	192)	147456	average_pooling2d_4[0
batch_normalization_31 (BatchNo	(None,	17,	17,	192)	576	conv2d_31[0][0]

batch_normalization_34 (BatchNo	(None,	17,	17,	192)	576	conv2d_34[0][0]
batch_normalization_39 (BatchNo	(None,	17,	17,	192)	576	conv2d_39[0][0]
batch_normalization_40 (BatchNo	(None,	17,	17,	192)	576	conv2d_40[0][0]
activation_31 (Activation) batch_normalization_31[0][0]	(None,	17,	17,	192)	0	
activation_34 (Activation) batch_normalization_34[0][0]	(None,	17,	17,	192)	0	
activation_39 (Activation) batch_normalization_39[0][0]	(None,	17,	17,	192)	0	
activation_40 (Activation) batch_normalization_40[0][0]	(None,	17,	17,	192)	0	
 mixed4 (Concatenate)	(None,	17,	17,	768)	0	activation_31[0][0] activation_34[0][0] activation_39[0][0] activation_40[0][0]
conv2d_45 (Conv2D)	(None,	17,	17,	160)	122880	mixed4[0][0]
batch_normalization_45 (BatchNo	(None,	17,	17,	160)	480	conv2d_45[0][0]
activation_45 (Activation) batch_normalization_45[0][0]	(None,	17,	17,	160)	0	
conv2d_46 (Conv2D)	(None,	17,	17,	160)	179200	activation_45[0][0]

batch_normalization_46 (BatchNo	(None,	17, 17	, 160)	480	conv2d_46[0][0]
activation_46 (Activation) batch_normalization_46[0][0]	(None,	17, 17	, 160)	0	
conv2d_42 (Conv2D)	(None,	17, 17	, 160)	122880	mixed4[0][0]
conv2d_47 (Conv2D)	(None,	17, 17	, 160)	179200	activation_46[0][0]
  batch_normalization_42 (BatchNo	(None,	17, 17	, 160)	480	conv2d_42[0][0]
  batch_normalization_47 (BatchNo	(None,	17, 17	, 160)	480	conv2d_47[0][0]
activation_42 (Activation) batch_normalization_42[0][0]	(None,	17, 17	, 160)	0	
activation_47 (Activation) batch_normalization_47[0][0]	(None,	17, 17	, 160)	0	
conv2d_43 (Conv2D)	(None,	17, 17	, 160)	179200	activation_42[0][0]
conv2d_48 (Conv2D)	(None,	17, 17	, 160)	179200	activation_47[0][0]
  batch_normalization_43 (BatchNo	(None,	17, 17	, 160)	480	conv2d_43[0][0]
  batch_normalization_48 (BatchNo	(None,	17, 17	, 160)	480	conv2d_48[0][0]
activation_43 (Activation) batch_normalization_43[0][0]	(None,	17, 17	, 160)	0	

activation_48 (Activation) batch_normalization_48[0][0]	(None,	17,	17,	160)	0	
average_pooling2d_5 (AveragePoo	(None,	17,	17,	768)	0	mixed4[0][0]
conv2d_41 (Conv2D)	(None,	17,	17,	192)	147456	mixed4[0][0]
conv2d_44 (Conv2D)	(None,	17,	17,	192)	215040	activation_43[0][0]
conv2d_49 (Conv2D)	(None,	17,	17,	192)	215040	activation_48[0][0]
 conv2d_50 (Conv2D) [0]	(None,	17,	17,	192)	147456	average_pooling2d_5[0
batch_normalization_41 (BatchNo	(None,	17,	17,	192)	576	conv2d_41[0][0]
batch_normalization_44 (BatchNo	(None,	17,	17,	192)	576	conv2d_44[0][0]
 batch_normalization_49 (BatchNo	(None,	17,	17,	192)	576	conv2d_49[0][0]
 batch_normalization_50 (BatchNo	(None,	17,	17,	192)	576	conv2d_50[0][0]
activation_41 (Activation) batch_normalization_41[0][0]	(None,	17,	17,	192)	0	
activation_44 (Activation) batch_normalization_44[0][0]	(None,	17,	17,	192)	0	
 activation_49 (Activation) batch_normalization_49[0][0]	(None,	17,	17,	192)	0	

activation_50 (Activation) batch_normalization_50[0][0]	(None,	17,	17,	192)	0	
 mixed5 (Concatenate)	(None,	17,	17,	768)	0	activation_41[0][0] activation_44[0][0] activation_49[0][0] activation_50[0][0]
conv2d_55(Conv2D)	(None,	17,	17,	160)	122880	mixed5[0][0]
 batch_normalization_55 (BatchNo	(None,	17,	17,	160)	480	conv2d_55[0][0]
activation_55 (Activation) batch_normalization_55[0][0]	(None,	17,	17,	160)	0	
 conv2d_56 (Conv2D)	(None,	17,	17,	160)	179200	activation_55[0][0]
batch_normalization_56 (BatchNo	(None,	17,	17,	160)	480	conv2d_56[0][0]
activation_56 (Activation) batch_normalization_56[0][0]	(None,	17,	17,	160)	0	
conv2d_52 (Conv2D)	(None,	17,	17,	160)	122880	mixed5[0][0]
 conv2d_57 (Conv2D)	(None,	17,	17,	160)	179200	activation_56[0][0]
  batch_normalization_52 (BatchNo	(None,	17,	17,	160)	480	conv2d_52[0][0]

batch_normalization_57 (BatchNo	(None,	17,	17,	160)	480	conv2d_57[0][0]
activation_52 (Activation) batch_normalization_52[0][0]	(None,	17,	17,	160)	0	
activation_57 (Activation) batch_normalization_57[0][0]	(None,	17,	17,	160)	0	
conv2d_53 (Conv2D)	(None,	17,	17,	160)	179200	activation_52[0][0]
conv2d_58(Conv2D)	(None,	17,	17,	160)	179200	activation_57[0][0]
batch_normalization_53 (BatchNo	(None,	17,	17,	160)	480	conv2d_53[0][0]
batch_normalization_58 (BatchNo	(None,	17,	17,	160)	480	conv2d_58[0][0]
activation_53 (Activation) batch_normalization_53[0][0]	(None,	17,	17,	160)	0	
activation_58 (Activation) batch_normalization_58[0][0]	(None,	17,	17,	160)	0	
average_pooling2d_6 (AveragePoo	(None,	17,	17,	768)	0	mixed5[0][0]
conv2d_51 (Conv2D)	(None,	17,	17,	192)	147456	mixed5[0][0]
conv2d_54 (Conv2D)	(None,	17,	17,	192)	215040	activation_53[0][0]
 	(None,	17,	17,	192)	215040	activation_58[0][0]

conv2d_60 (Conv2D) [0]	(None,	17,	17,	192)	147456	average_pooling2d_6[0]
batch_normalization_51 (BatchNo	(None,	17,	17,	192)	576	conv2d_51[0][0]
batch_normalization_54 (BatchNo	(None,	17,	17,	192)	576	conv2d_54[0][0]
batch_normalization_59 (BatchNo	(None,	17,	17,	192)	576	conv2d_59[0][0]
batch_normalization_60 (BatchNo	(None,	17,	17,	192)	576	conv2d_60[0][0]
activation_51 (Activation) batch_normalization_51[0][0]	(None,	17,	17,	192)	0	
activation_54 (Activation) batch_normalization_54[0][0]	(None,	17,	17,	192)	0	
activation_59 (Activation) batch_normalization_59[0][0]	(None,	17,	17,	192)	0	
activation_60 (Activation) batch_normalization_60[0][0]	(None,	17,	17,	192)	0	
mixed6 (Concatenate)	(None,	17,	17,	768)	0	activation_51[0][0] activation_54[0][0] activation_59[0][0] activation_60[0][0]
conv2d_65 (Conv2D)	(None,	17,	17,	192)	147456	mixed6[0][0]
  batch_normalization_65 (BatchNo	(None,	17,	17,	192)	576	conv2d_65[0][0]

(None,	17,	17,	192)	0	
(None,	17,	17,	192)	258048	activation_65[0][0]
(None,	17,	17,	192)	576	conv2d_66[0][0]
(None,	17,	17,	192)	0	
(None,	17,	17,	192)	147456	mixed6[0][0]
(None,	17,	17,	192)	258048	activation_66[0][0]
(None,	17,	17,	192)	576	conv2d_62[0][0]
(None,	17,	17,	192)	576	conv2d_67[0][0]
(None,	17,	17,	192)	0	
(None,	17,	17,	192)	0	
(None,	17,	17,	192)	258048	activation_62[0][0]
(None,	17,	17,	192)	258048	activation_67[0][0]
(None,	17,	17,	192)	576	conv2d_63[0][0]
	(None,  (None,  (None,  (None,  (None,  (None,  (None,  (None,  (None,	(None, 17,  (None, 17,	(None, 17, 17,  (None, 17, 17,	(None, 17, 17, 192)  (None, 17, 17, 192)	(None, 17, 17, 192) 0  (None, 17, 17, 192) 258048  (None, 17, 17, 192) 576  (None, 17, 17, 192) 147456  (None, 17, 17, 192) 258048  (None, 17, 17, 192) 576  (None, 17, 17, 192) 576  (None, 17, 17, 192) 0  (None, 17, 17, 192) 0  (None, 17, 17, 192) 258048  (None, 17, 17, 192) 258048  (None, 17, 17, 192) 576

batch_normalization_68 (BatchNo	(None,	17,	17,	192)	576	conv2d_68[0][0]
activation_63 (Activation) batch_normalization_63[0][0]	(None,	17,	17,	192)	0	
activation_68 (Activation) batch_normalization_68[0][0]	(None,	17,	17,	192)	0	
average_pooling2d_7 (AveragePoo	(None,	17,	17,	768)	0	mixed6[0][0]
conv2d_61 (Conv2D)	(None,	17,	17,	192)	147456	mixed6[0][0]
conv2d_64 (Conv2D)	(None,	17,	17,	192)	258048	activation_63[0][0]
conv2d_69 (Conv2D)	(None,	17,	17,	192)	258048	activation_68[0][0]
 conv2d_70 (Conv2D) [0]	(None,	17,	17,	192)	147456	average_pooling2d_7[0]
  batch_normalization_61 (BatchNo	(None,	17,	17,	192)	576	conv2d_61[0][0]
batch_normalization_64 (BatchNo	(None,	17,	17,	192)	576	conv2d_64[0][0]
 batch_normalization_69 (BatchNo	(None,	17,	17,	192)	576	conv2d_69[0][0]
batch_normalization_70 (BatchNo	(None,	17,	17,	192)	576	conv2d_70[0][0]
activation_61 (Activation) batch_normalization_61[0][0]	(None,	17,	17,	192)	0	

activation_64 (Activation) batch_normalization_64[0][0]	(None,	17,	17,	192)	0	
activation_69 (Activation) batch_normalization_69[0][0]	(None,	17,	17,	192)	0	
activation_70 (Activation) batch_normalization_70[0][0]	(None,	17,	17,	192)	0	
 mixed7 (Concatenate)	(None,	17,	17,	768)	0	activation_61[0][0] activation_64[0][0] activation_69[0][0] activation_70[0][0]
conv2d_73 (Conv2D)	(None,	17,	17,	192)	147456	mixed7[0][0]
  batch_normalization_73 (BatchNo	(None,	17,	17,	192)	576	conv2d_73[0][0]
activation_73 (Activation) batch_normalization_73[0][0]	(None,	17,	17,	192)	0	
conv2d_74(Conv2D)	(None,	17,	17,	192)	258048	activation_73[0][0]
 batch_normalization_74 (BatchNo	(None,	17,	17,	192)	576	conv2d_74[0][0]
activation_74 (Activation) batch_normalization_74[0][0]	(None,	17,	17,	192)	0	
					147456	mixed7[0][0]

conv2d_75 (Conv2D)	(None,	17, 17, 192)	258048	activation_74[0][0]
batch_normalization_71 (BatchNo	(None,	17, 17, 192)	576	conv2d_71[0][0]
batch_normalization_75 (BatchNo	(None,	17, 17, 192)	576	conv2d_75[0][0]
activation_71 (Activation) batch_normalization_71[0][0]	(None,	17, 17, 192)	0	
activation_75 (Activation) batch_normalization_75[0][0]	(None,	17, 17, 192)	0	
conv2d_72(Conv2D)	(None,	8, 8, 320)	552960	activation_71[0][0]
conv2d_76 (Conv2D)	(None,	8, 8, 192)	331776	activation_75[0][0]
batch_normalization_72 (BatchNo	(None,	8, 8, 320)	960	conv2d_72[0][0]
batch_normalization_76 (BatchNo	(None,	8, 8, 192)	576	conv2d_76[0][0]
activation_72 (Activation) batch_normalization_72[0][0]	(None,	8, 8, 320)	0	
activation_76 (Activation) batch_normalization_76[0][0]	(None,	8, 8, 192)	0	
max_pooling2d_4 (MaxPooling2D)	(None,	8, 8, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None,	8, 8, 1280)	0	activation_72[0][0] activation_76[0][0]

						max_pooling2d_4[0][
conv2d_81 (Conv2D)	(None,	8,	8,	448)	573440	mixed8[0][0]
batch_normalization_81 (BatchNo	(None,	8,	8,	448)	1344	conv2d_81[0][0]
activation_81 (Activation) batch_normalization_81[0][0]	(None,	8,	8,	448)	0	
conv2d_78(Conv2D)	(None,	8,	8,	384)	491520	mixed8[0][0]
conv2d_82 (Conv2D)	(None,	8,	8,	384)	1548288	activation_81[0][0]
batch_normalization_78 (BatchNo	(None,	8,	8,	384)	1152	conv2d_78[0][0]
batch_normalization_82 (BatchNo	(None,	8,	8,	384)	1152	conv2d_82[0][0]
activation_78 (Activation) batch_normalization_78[0][0]	(None,	8,	8,	384)	0	
activation_82 (Activation) batch_normalization_82[0][0]	(None,	8,	8,	384)	0	
conv2d_79 (Conv2D)	(None,	8,	8,	384)	442368	activation_78[0][0]
conv2d_80 (Conv2D)	(None,	8,	8,	384)	442368	activation_78[0][0]
  conv2d_83 (Conv2D)	(None,	8,	8,	384)	442368	activation_82[0][0]

conv2d_84 (Conv2D)	(None,	8,	8,	384)	442368	activation_82[0][0]
average_pooling2d_8 (AveragePoo	(None,	8,	8,	1280)	0	mixed8[0][0]
conv2d_77 (Conv2D)	(None,	8,	8,	320)	409600	mixed8[0][0]
batch_normalization_79 (BatchNo	(None,	8,	8,	384)	1152	conv2d_79[0][0]
batch_normalization_80 (BatchNo	(None,	8,	8,	384)	1152	conv2d_80[0][0]
batch_normalization_83 (BatchNo	(None,	8,	8,	384)	1152	conv2d_83[0][0]
  batch_normalization_84 (BatchNo	(None,	8,	8,	384)	1152	conv2d_84[0][0]
 conv2d_85 (Conv2D) [0]	(None,	8,	8,	192)	245760	average_pooling2d_8[0
batch_normalization_77 (BatchNo	(None,	8,	8,	320)	960	conv2d_77[0][0]
activation_79 (Activation) batch_normalization_79[0][0]	(None,	8,	8,	384)	0	
activation_80 (Activation) batch_normalization_80[0][0]	(None,	8,	8,	384)	0	
activation_83 (Activation) batch_normalization_83[0][0]	(None,	8,	8,	384)	0	
activation_84 (Activation) batch_normalization_84[0][0]	(None,	8,	8,	384)	0	

batch_normalization_85 (BatchNo	(None,	8,	8,	192)	576	conv2d_85[0][0]
activation_77 (Activation) batch_normalization_77[0][0]	(None,	8,	8,	320)	0	
 mixed9_0 (Concatenate)	(None,	8,	8,	768)	0	activation_79[0][0] activation_80[0][0]
 concatenate_1 (Concatenate)	(None,	8,	8,	768)	0	activation_83[0][0] activation_84[0][0]
activation_85 (Activation) batch_normalization_85[0][0]	(None,	8,	8,	192)	0	
 mixed9 (Concatenate)	(None,	8,	8,	2048)	0	activation_77[0][0]  mixed9_0[0][0]  concatenate_1[0][0]  activation_85[0][0]
conv2d_90 (Conv2D)	(None,	8,	8,	448)	917504	mixed9[0][0]
batch_normalization_90 (BatchNo	(None,	8,	8,	448)	1344	conv2d_90[0][0]
activation_90 (Activation) batch_normalization_90[0][0]	(None,	8,	8,	448)	0	
 conv2d_87 (Conv2D)	(None,	8,	8,	384)	786432	mixed9[0][0]
conv2d_91 (Conv2D)	(None,	8,	8,	384)	1548288	activation_90[0][0]

batch_normalization_87 (BatchNo	(None,	8,	8,	384)	1152	conv2d_87[0][0]
batch_normalization_91 (BatchNo	(None,	8,	8,	384)	1152	conv2d_91[0][0]
activation_87 (Activation) batch_normalization_87[0][0]	(None,	8,	8,	384)	0	
activation_91 (Activation) batch_normalization_91[0][0]	(None,	8,	8,	384)	0	
conv2d_88 (Conv2D)	(None,	8,	8,	384)	442368	activation_87[0][0]
conv2d_89(Conv2D)	(None,	8,	8,	384)	442368	activation_87[0][0]
conv2d_92 (Conv2D)	(None,	8,	8,	384)	442368	activation_91[0][0]
conv2d_93 (Conv2D)	(None,	8,	8,	384)	442368	activation_91[0][0]
average_pooling2d_9 (AveragePoo	(None,	8,	8,	2048)	0	mixed9[0][0]
conv2d_86 (Conv2D)	(None,	8,	8,	320)	655360	mixed9[0][0]
batch_normalization_88 (BatchNo	(None,	8,	8,	384)	1152	conv2d_88[0][0]
batch_normalization_89 (BatchNo	(None,	8,	8,	384)	1152	conv2d_89[0][0]
batch_normalization_92 (BatchNo	(None,	8,	8,	384)	1152	conv2d_92[0][0]

batch_normalization_93 (BatchNo	(None,	8,	8,	384)	1152	conv2d_93[0][0]
	(None,	8,	8,	192)	393216	average_pooling2d_9[0]
batch_normalization_86 (BatchNo	(None,	8,	8,	320)	960	conv2d_86[0][0]
activation_88 (Activation) batch_normalization_88[0][0]	(None,	8,	8,	384)	0	
activation_89 (Activation) batch_normalization_89[0][0]	(None,	8,	8,	384)	0	
activation_92 (Activation) batch_normalization_92[0][0]	(None,	8,	8,	384)	0	
activation_93 (Activation) batch_normalization_93[0][0]	(None,	8,	8,	384)	0	
 batch_normalization_94 (BatchNo	(None,	8,	8,	192)	576	conv2d_94[0][0]
activation_86 (Activation) batch_normalization_86[0][0]	(None,	8,	8,	320)	0	
mixed9_1 (Concatenate)	(None,	8,	8,	768)	0	activation_88[0][0] activation_89[0][0]
concatenate_2 (Concatenate)	(None,	8,	8,	768)	0	activation_92[0][0] activation_93[0][0]
activation_94 (Activation) batch_normalization_94[0][0]	(None,	8,	8,	192)	0	

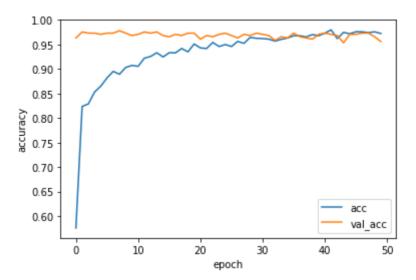
```
mixed10 (Concatenate)
                        (None, 8, 8, 2048) 0
                                                    activation_86[0][0]
                                                    mixed9_1[0][0]
                                                    concatenate_2[0][0]
                                                    activation_94[0][0]
global_average_pooling2d_1 (Glo (None, 2048)
                                                    mixed10[0][0]
dense_1 (Dense)
                         (None, 512)
                                         1049088
global_average_pooling2d_1[0][0]
dense_2 (Dense)
                                          8208
                                                    dense_1[0][0]
                          (None, 16)
______
Total params: 22,860,080
Trainable params: 1,057,296
Non-trainable params: 21,802,784
```

```
Epoch 1/50
- 39s - loss: 1.6997 - acc: 0.5754 - val_loss: 0.4074 - val_acc: 0.9632
Epoch 2/50
- 34s - loss: 0.7656 - acc: 0.8229 - val_loss: 0.1433 - val_acc: 0.9755
Epoch 3/50
- 33s - loss: 0.6031 - acc: 0.8257 - val_loss: 0.1204 - val_acc: 0.9730
Epoch 4/50
- 33s - loss: 0.5218 - acc: 0.8543 - val_loss: 0.1181 - val_acc: 0.9730
Epoch 5/50
- 33s - loss: 0.4749 - acc: 0.8663 - val_loss: 0.1043 - val_acc: 0.9706
Epoch 6/50
- 33s - loss: 0.4203 - acc: 0.8823 - val_loss: 0.0980 - val_acc: 0.9730
Epoch 7/50
```

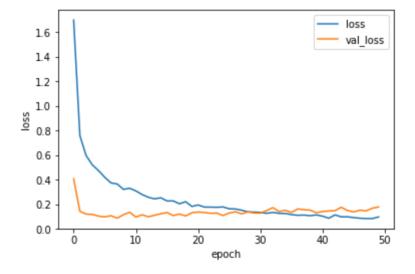
```
- 33s - loss: 0.3769 - acc: 0.8960 - val_loss: 0.1073 - val_acc: 0.9730
Epoch 8/50
 - 33s - loss: 0.3645 - acc: 0.8903 - val_loss: 0.0876 - val_acc: 0.9779
Epoch 9/50
- 32s - loss: 0.3208 - acc: 0.9040 - val_loss: 0.1158 - val_acc: 0.9730
Epoch 10/50
 - 32s - loss: 0.3314 - acc: 0.9080 - val_loss: 0.1357 - val_acc: 0.9681
Epoch 11/50
- 32s - loss: 0.3078 - acc: 0.9063 - val_loss: 0.0965 - val_acc: 0.9706
Epoch 12/50
- 32s - loss: 0.2801 - acc: 0.9223 - val_loss: 0.1151 - val_acc: 0.9755
Epoch 13/50
- 32s - loss: 0.2574 - acc: 0.9263 - val_loss: 0.0987 - val_acc: 0.9730
Epoch 14/50
- 32s - loss: 0.2439 - acc: 0.9337 - val_loss: 0.1115 - val_acc: 0.9755
Epoch 15/50
- 32s - loss: 0.2537 - acc: 0.9251 - val_loss: 0.1231 - val_acc: 0.9681
Epoch 16/50
- 32s - loss: 0.2262 - acc: 0.9337 - val_loss: 0.1333 - val_acc: 0.9657
Epoch 17/50
- 32s - loss: 0.2281 - acc: 0.9331 - val_loss: 0.1087 - val_acc: 0.9706
Epoch 18/50
 - 32s - loss: 0.2047 - acc: 0.9423 - val_loss: 0.1209 - val_acc: 0.9681
Epoch 19/50
- 32s - loss: 0.2206 - acc: 0.9354 - val_loss: 0.1060 - val_acc: 0.9730
Epoch 20/50
- 33s - loss: 0.1826 - acc: 0.9514 - val_loss: 0.1315 - val_acc: 0.9730
Epoch 21/50
- 34s - loss: 0.1942 - acc: 0.9434 - val_loss: 0.1375 - val_acc: 0.9608
Epoch 22/50
- 32s - loss: 0.1767 - acc: 0.9423 - val_loss: 0.1343 - val_acc: 0.9681
Epoch 23/50
- 32s - loss: 0.1769 - acc: 0.9543 - val_loss: 0.1268 - val_acc: 0.9657
Epoch 24/50
- 32s - loss: 0.1750 - acc: 0.9463 - val_loss: 0.1292 - val_acc: 0.9706
Epoch 25/50
- 32s - loss: 0.1777 - acc: 0.9503 - val_loss: 0.1086 - val_acc: 0.9730
Epoch 26/50
- 32s - loss: 0.1630 - acc: 0.9463 - val_loss: 0.1297 - val_acc: 0.9681
Epoch 27/50
- 33s - loss: 0.1614 - acc: 0.9566 - val_loss: 0.1394 - val_acc: 0.9632
Epoch 28/50
- 31s - loss: 0.1515 - acc: 0.9526 - val_loss: 0.1221 - val_acc: 0.9706
Epoch 29/50
- 31s - loss: 0.1386 - acc: 0.9646 - val_loss: 0.1387 - val_acc: 0.9681
Epoch 30/50
 - 31s - loss: 0.1369 - acc: 0.9629 - val_loss: 0.1295 - val_acc: 0.9730
Epoch 31/50
- 31s - loss: 0.1341 - acc: 0.9623 - val_loss: 0.1286 - val_acc: 0.9706
Epoch 32/50
- 31s - loss: 0.1268 - acc: 0.9611 - val_loss: 0.1489 - val_acc: 0.9681
Epoch 33/50
 - 31s - loss: 0.1329 - acc: 0.9571 - val_loss: 0.1732 - val_acc: 0.9583
```

```
Epoch 34/50
- 31s - loss: 0.1268 - acc: 0.9606 - val_loss: 0.1412 - val_acc: 0.9657
Epoch 35/50
- 31s - loss: 0.1249 - acc: 0.9634 - val_loss: 0.1515 - val_acc: 0.9632
Epoch 36/50
- 31s - loss: 0.1162 - acc: 0.9680 - val_loss: 0.1339 - val_acc: 0.9730
Epoch 37/50
- 32s - loss: 0.1106 - acc: 0.9680 - val_loss: 0.1624 - val_acc: 0.9657
Epoch 38/50
- 32s - loss: 0.1118 - acc: 0.9657 - val_loss: 0.1561 - val_acc: 0.9632
Epoch 39/50
- 32s - loss: 0.1071 - acc: 0.9703 - val_loss: 0.1529 - val_acc: 0.9608
Epoch 40/50
- 31s - loss: 0.1135 - acc: 0.9680 - val_loss: 0.1315 - val_acc: 0.9706
Epoch 41/50
- 32s - loss: 0.1033 - acc: 0.9731 - val_loss: 0.1418 - val_acc: 0.9730
Epoch 42/50
- 33s - loss: 0.0868 - acc: 0.9800 - val_loss: 0.1460 - val_acc: 0.9706
Epoch 43/50
- 33s - loss: 0.1137 - acc: 0.9623 - val_loss: 0.1483 - val_acc: 0.9681
Epoch 44/50
- 32s - loss: 0.0982 - acc: 0.9749 - val_loss: 0.1763 - val_acc: 0.9534
Epoch 45/50
- 31s - loss: 0.0980 - acc: 0.9720 - val_loss: 0.1496 - val_acc: 0.9706
Epoch 46/50
- 33s - loss: 0.0917 - acc: 0.9760 - val_loss: 0.1386 - val_acc: 0.9706
Epoch 47/50
- 32s - loss: 0.0874 - acc: 0.9760 - val_loss: 0.1525 - val_acc: 0.9730
Epoch 48/50
- 32s - loss: 0.0838 - acc: 0.9743 - val_loss: 0.1467 - val_acc: 0.9730
Epoch 49/50
- 32s - loss: 0.0839 - acc: 0.9760 - val_loss: 0.1689 - val_acc: 0.9657
Epoch 50/50
- 32s - loss: 0.0982 - acc: 0.9726 - val_loss: 0.1786 - val_acc: 0.9559
```

```
plt.plot(history.history['acc'], label='acc')
plt.plot(history.history['val_acc'], label='val_acc')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(loc='best')
plt.show()
```



```
plt.plot(history.history['loss'], label='loss')
plt.plot(history.history['val_loss'], label='val_loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(loc='best')
plt.show()
```



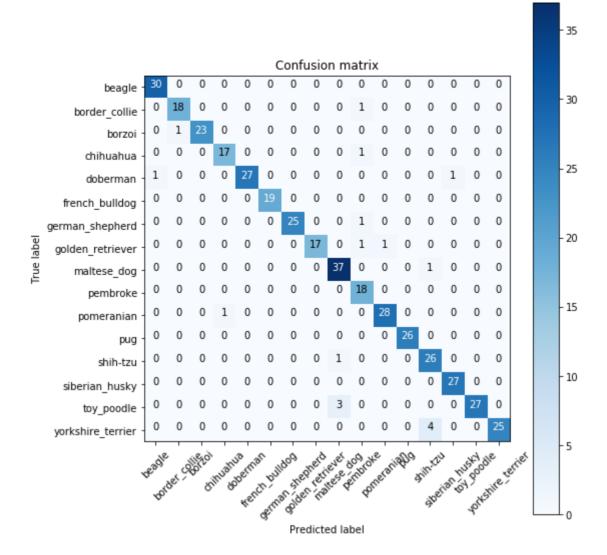
```
score = model.evaluate(X_test, y_test, batch_size=32)
list(zip(model.metrics_names, score))
```

```
[('loss', 0.17864525814851126), ('acc', 0.955882351772458)]
```

```
from pylab import rcParams
rcParams['figure.figsize'] = (8, 8)
def plot_confusion_matrix(cm, breeds, normalize=False, title='Confusion matrix',
    cmap=plt.cm.Blues):
   This function prints and plots the confusion matrix.
   Normalization can be applied by setting `normalize=True`.
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
       print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(breeds))
    plt.xticks(tick_marks, breeds, rotation=45)
    plt.yticks(tick_marks, breeds)
    fmt = '.2f' if normalize else 'd'
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, format(cm[i, j], fmt),
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")
    plt.tight_layout()
    plt.xlabel('Predicted label')
    plt.ylabel('True label')
    plt.show()
```

```
y_pred = model.predict(x_test)
y_pred = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_test, axis=1)
confusion_mtx = confusion_matrix(y_true, y_pred)
plot_confusion_matrix(confusion_mtx, breeds)
```

Confusion matrix, without normalization

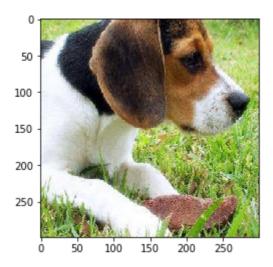


```
test_index = 10
rcParams['figure.figsize'] = (4, 4)

plt.imshow(X[test_index, :, :, :])
test_img = np.reshape(X[test_index, :, :, :], (1, INPUT_SIZE, INPUT_SIZE, 3))

predict = model.predict(test_img)
print(breeds[np.argmax(predict)])
```

beagle

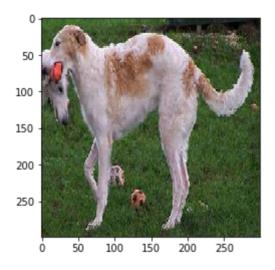


```
test_index = 200
rcParams['figure.figsize'] = (4, 4)

plt.imshow(X[test_index, :, :, :])
test_img = np.reshape(X[test_index, :, :, :], (1, INPUT_SIZE, INPUT_SIZE, 3))

predict = model.predict(test_img)
print(breeds[np.argmax(predict)])
```

borzoi



```
model.save('model_16.h5')
```