

NAME : AAKASH MOHAN
ID : 1001656408

SUBJECT CODE : EE5353

SUBJECT : Neural Networks and Deep Learning

PROGRAM ASSIGNMENT 9:

Transfer Learning using Google Colab

```

import tensorflow as tf
import random as rn
import os,cv2
import numpy as np

os.environ['PYTHONHASHSEED'] = '0'
# Setting the seed for numpy-generated random numbers
np.random.seed(37)
# Setting the seed for python random numbers
rn.seed(1254)
# Setting the graph-level random seed.
tf.set_random_seed(89)
from keras import backend as K
session_conf = tf.ConfigProto(intra_op_parallelism_threads=1,inter_op_parallelism_threads=

#Force Tensorflow to use a single thread
sess = tf.Session(graph=tf.get_default_graph(), config=session_conf)
K.set_session(sess)
import glob

from sklearn.utils import shuffle

from sklearn.model_selection import train_test_split

import re
from keras.models import Sequential
from keras import models
from keras import layers
from keras import optimizers
from keras.applications import VGG16
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from keras.layers import Dense, Conv2D, Flatten, MaxPooling2D, Dropout
from sklearn.utils import shuffle
from PIL import Image
from google.colab.patches import cv2_imshow
import matplotlib.pyplot as plt
import glob

def gen_image(img):
    plt.imshow(img)
    return plt

#from sklearn.cross_validation import train_test_split

from google.colab import drive
drive.mount('/content/drive')

PATH = os.getcwd()
from google.colab import drive
drive.mount('/content/drive')

```

```

train_dir = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/train'
validation_dir = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/validatio

#create model
model = Sequential()
#add model layers

model.add(Conv2D(128, kernel_size=3, strides=1, activation='relu', input_shape=(256, 256, 3)))
    # 64 are the number of filters, kernel size is the size of the filters example 3*3
model.add(Conv2D(128, kernel_size=3, strides=1, activation='relu'))
model.add(MaxPooling2D(pool_size=(3, 3)))
model.add(Conv2D(256, kernel_size=3, strides=1, activation='relu'))
model.add(Conv2D(256, kernel_size=3, strides=1, activation='relu'))
model.add(Flatten())
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(4, activation='softmax'))

image_size=256
train_datagen = ImageDataGenerator(
    rescale=1./255,
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest')

validation_datagen = ImageDataGenerator(rescale=1./255)

# Change the batchsize according to your system RAM
train_batchsize = 50
val_batchsize = 10

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(image_size, image_size),
    batch_size=train_batchsize,
    class_mode='categorical')

validation_generator = validation_datagen.flow_from_directory(
    validation_dir,
    target_size=(image_size, image_size),
    batch_size=val_batchsize,
    class_mode='categorical',
    shuffle=False)

# Compile the model
model.compile(loss='categorical_crossentropy',
              optimizer=optimizers.RMSprop(lr=1e-4),
              metrics=['acc'])

```

```

# Train the model
history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.samples/train_generator.batch_size ,
    epochs=5,
    validation_data=validation_generator,
    validation_steps=validation_generator.samples/validation_generator.batch_size,
    verbose=1)

'''#TESTING
test_dir='/content/drive/My Drive/images'
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory( test_dir, target_size=(256, 256),batch_
test_generator=test_generator.reshape(10,256,256,3)

ytested = model.predict_classes(test_generator)
labels=['human','horse','dogs','cats']
for i in range(len(ytested)):
    #print("The Predicted Testing image is =%s verify below" ,labels[ytested[i]])
    print(np.argmax(ytested[i]))
    gen_image(test_generator[i]).show() '''

#TESTING
data_path = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/test/all_data'
img_data_list=[]
img_list = glob.glob(data_path+'/*.jpg')
for img in img_list:
    input_img=cv2.imread(img,1 )
    input_img_resize=cv2.resize(input_img,(256,256))
    img_data_list.append(input_img_resize)

img_data = np.array(img_data_list)
img_data = img_data.astype('float32')
x_test = shuffle(img_data, random_state=2)
x_test=x_test/255
#Nv_test=x_test.shape[0]
x_test = x_test.reshape(40,256,256,3)
label=['humans','horse','dogs','cats']
ytested = model.predict_classes(x_test)
for i in range(40):
    print("The Predicted Testing image is =%s verify below" %label[ytested[i]])
    gen_image(x_test[i]).show() # printing image vs the predicted image below

```



The default version of TensorFlow in Colab will soon switch to TensorFlow 2.x.
We recommend you [upgrade](#) now or ensure your notebook will continue to use TensorFlow 1.x via the %ten

Using TensorFlow backend.

Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client_id=94731

Enter your authorization code:

.....

Mounted at /content/drive

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mo

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

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WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

Instructions for updating:

Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`

Found 608 images belonging to 4 classes.

Found 160 images belonging to 4 classes.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

Epoch 1/5

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

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WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflo

13/12 [=====] - 221s 17s/step - loss: 1.7662 - acc: 0.2818

Epoch 2/5

13/12 [=====] - 7s 546ms/step - loss: 1.3769 - acc: 0.2716

Epoch 3/5

13/12 [=====] - 13s 1s/step - loss: 1.3905 - acc: 0.2853 -

Epoch 4/5

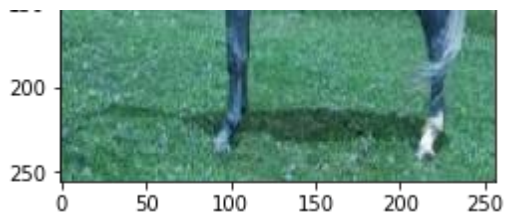
13/12 [=====] - 13s 1s/step - loss: 1.3920 - acc: 0.2856 -

Epoch 5/5

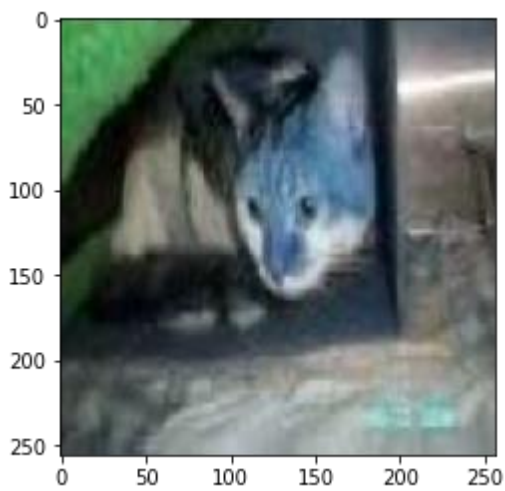
13/12 [=====] - 13s 1s/step - loss: 1.3838 - acc: 0.2796 -

The Predicted Testing image is =humans verify below

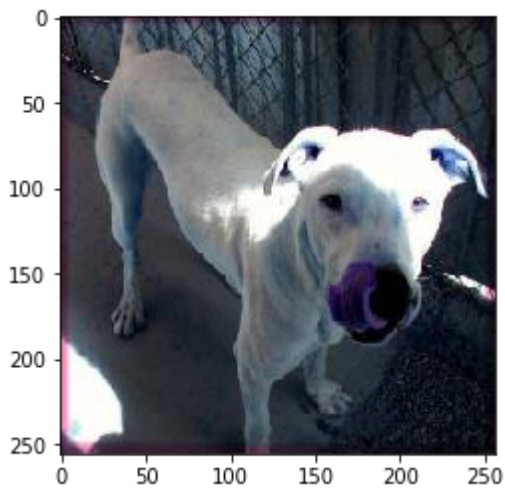




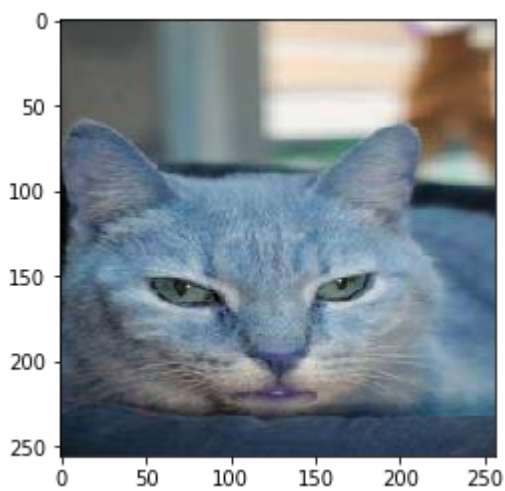
The Predicted Testing image is =humans verify below



The Predicted Testing image is =humans verify below

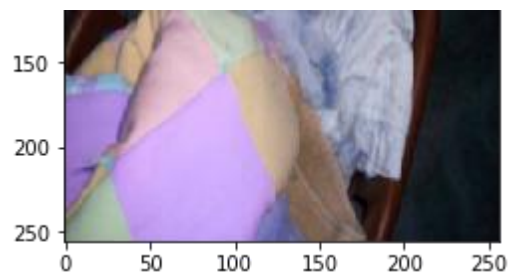


The Predicted Testing image is =horse verify below

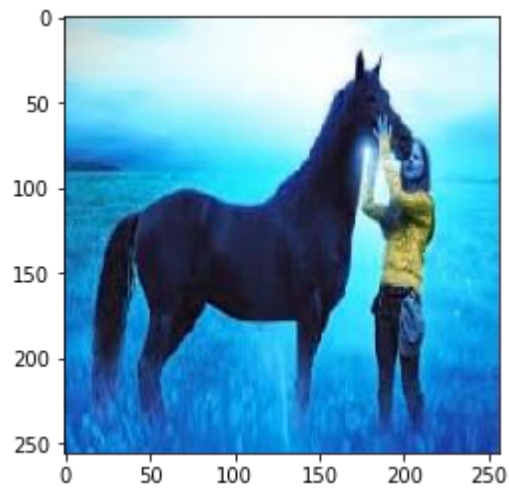


The Predicted Testing image is =humans verify below

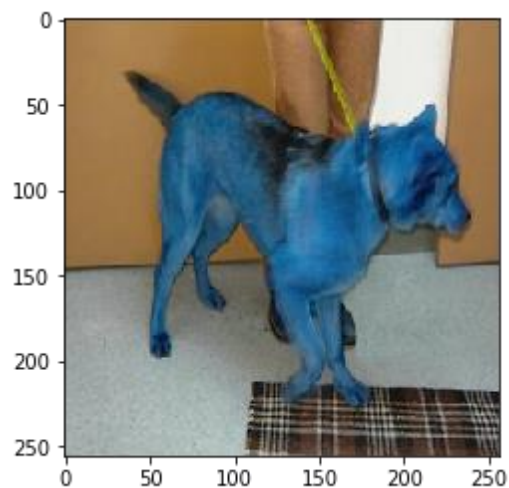




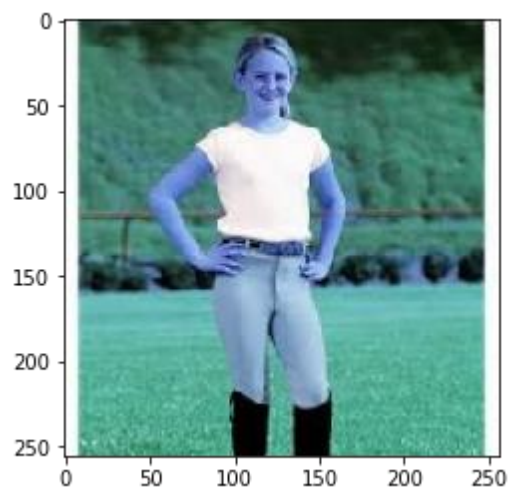
The Predicted Testing image is =humans verify below



The Predicted Testing image is =cats verify below

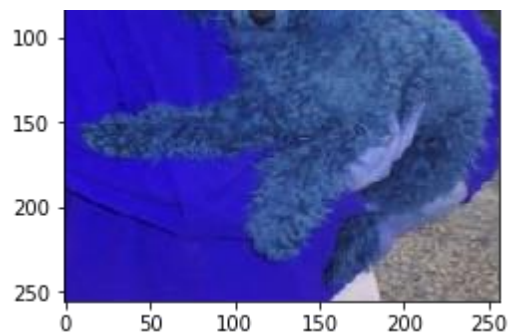


The Predicted Testing image is =humans verify below

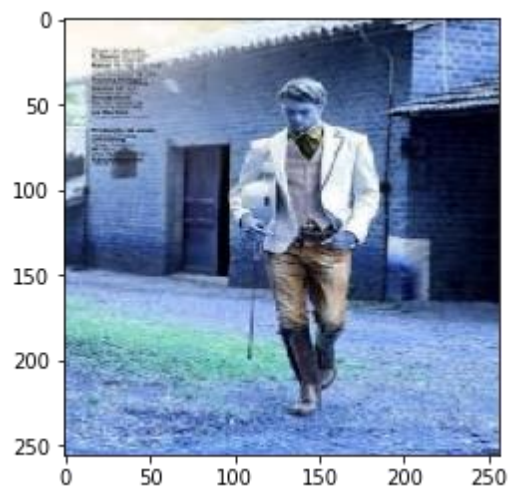


The Predicted Testing image is =humans verify below

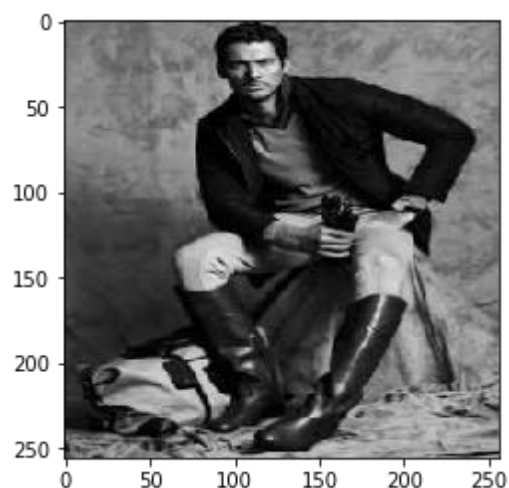




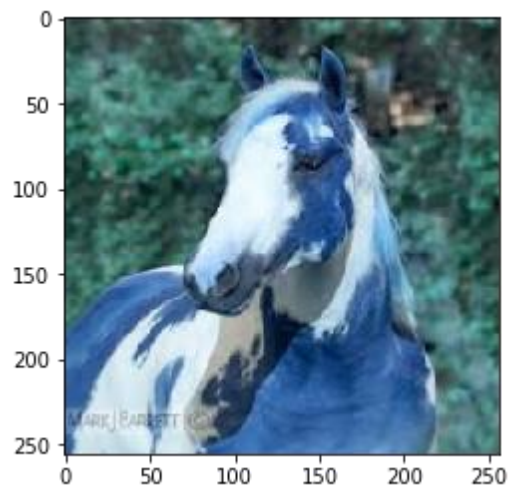
The Predicted Testing image is =humans verify below



The Predicted Testing image is =humans verify below

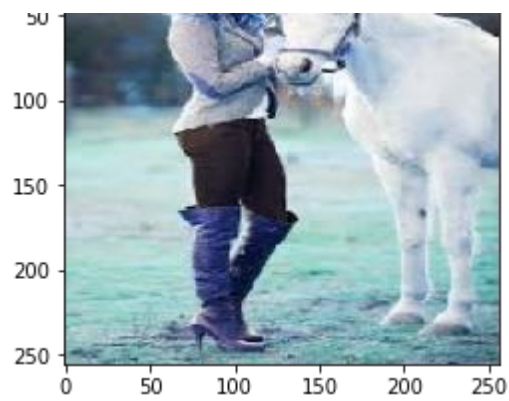


The Predicted Testing image is =humans verify below

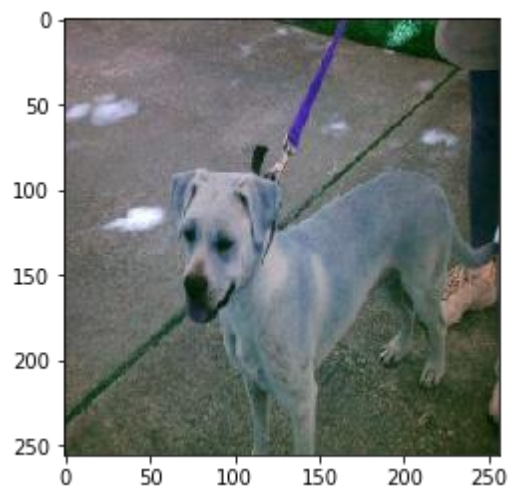


The Predicted Testing image is =humans verify below

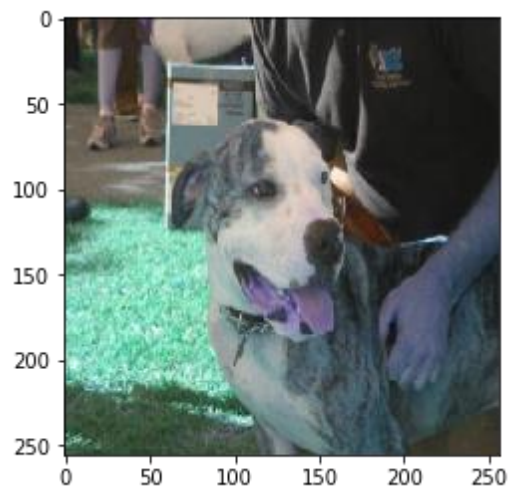




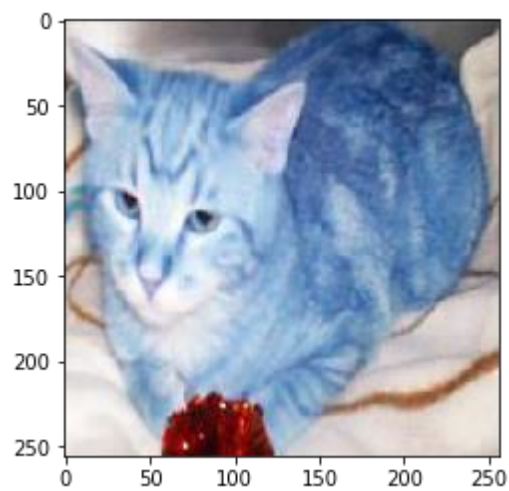
The Predicted Testing image is =cats verify below



The Predicted Testing image is =humans verify below

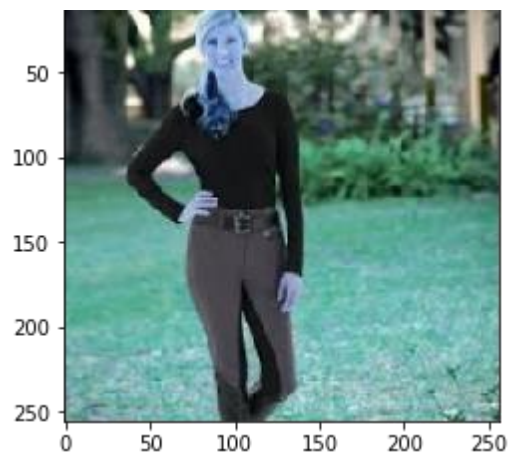


The Predicted Testing image is =cats verify below

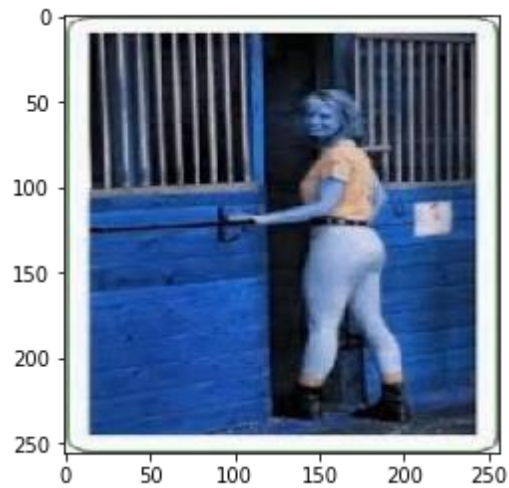


The Predicted Testing image is =humans verify below

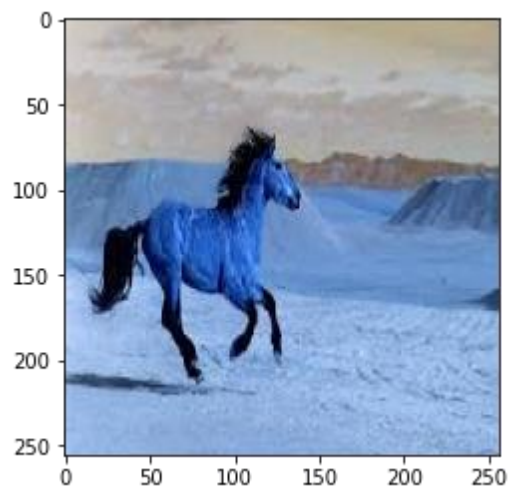




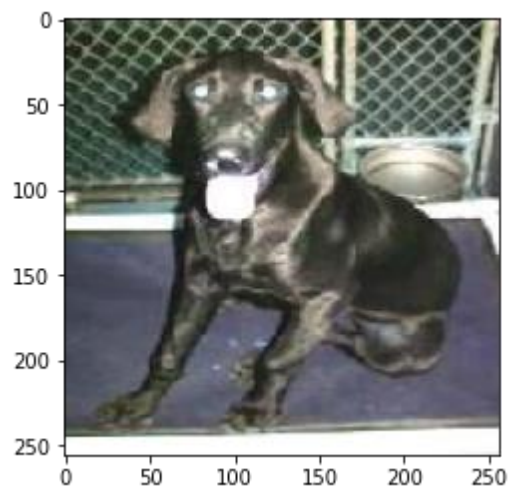
The Predicted Testing image is =humans verify below



The Predicted Testing image is =humans verify below



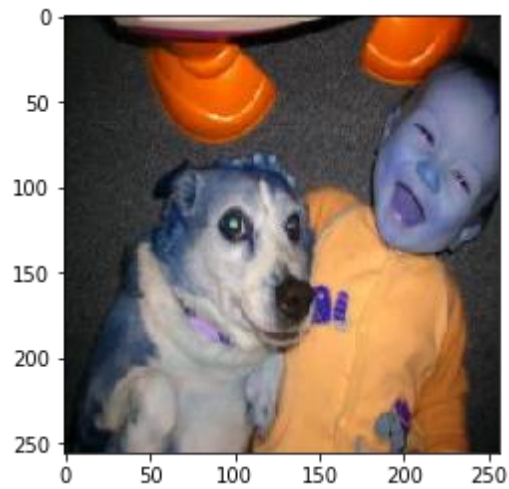
The Predicted Testing image is =humans verify below



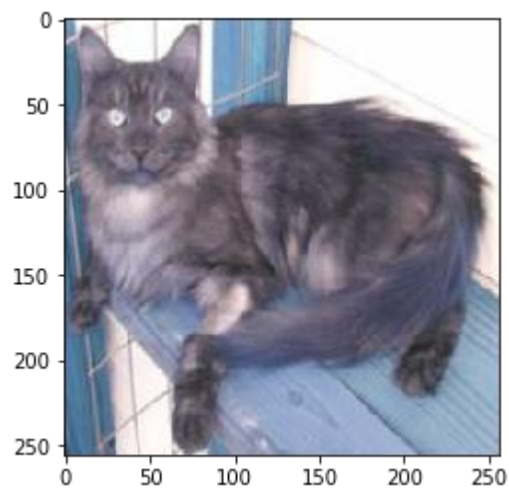
he Predicted Testing image is =humans verify below



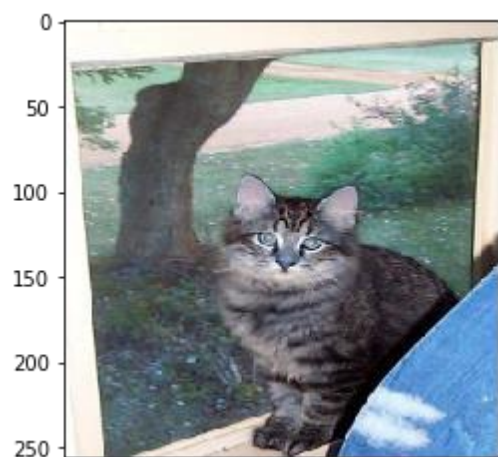
The Predicted Testing image is =cats verify below



The Predicted Testing image is =cats verify below

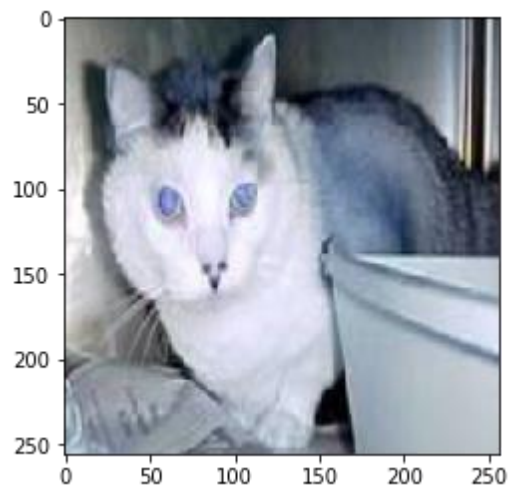


The Predicted Testing image is =humans verify below

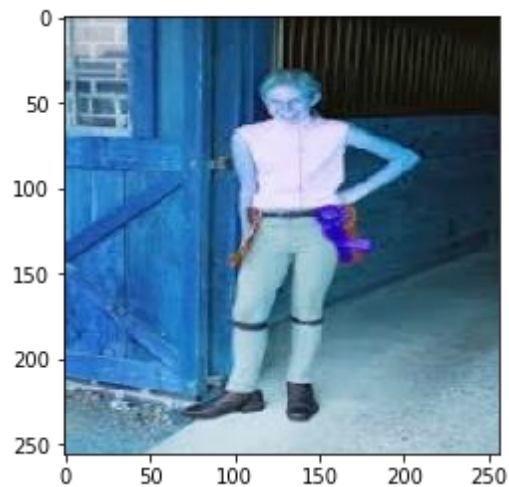


0 50 100 150 200 250

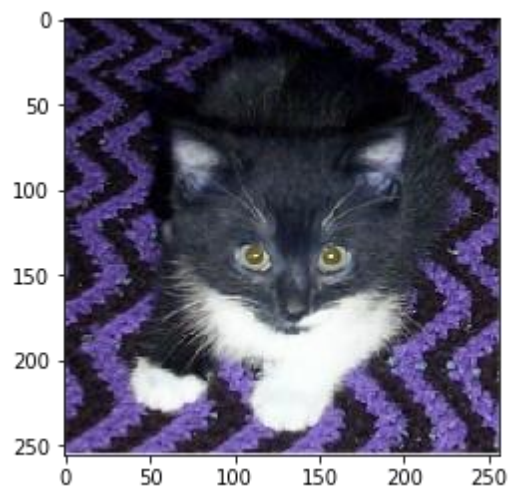
The Predicted Testing image is =humans verify below



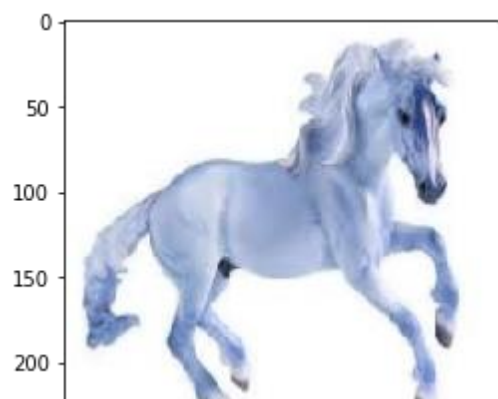
The Predicted Testing image is =humans verify below

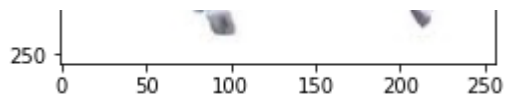


The Predicted Testing image is =humans verify below

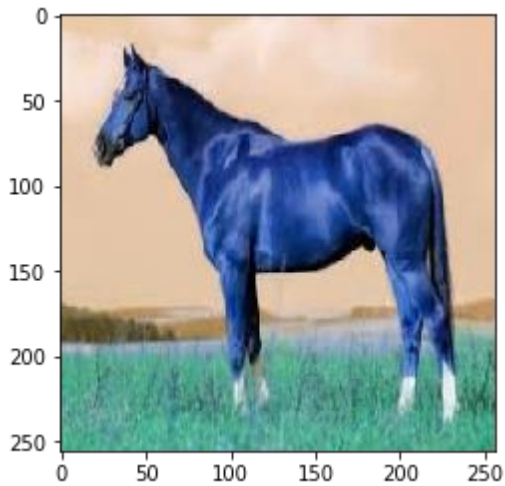


The Predicted Testing image is =cats verify below

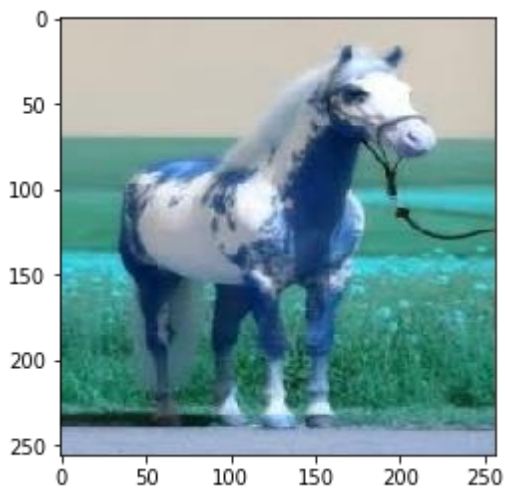




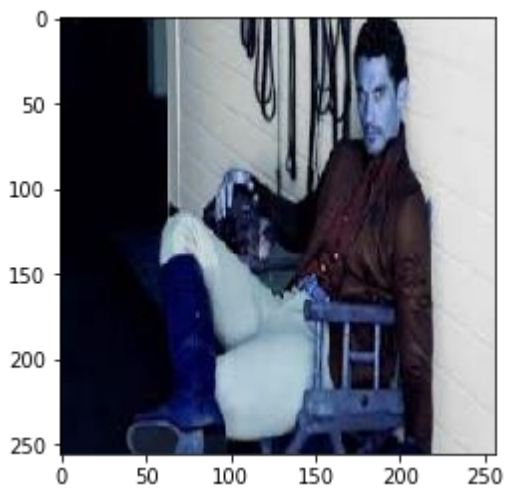
The Predicted Testing image is =humans verify below



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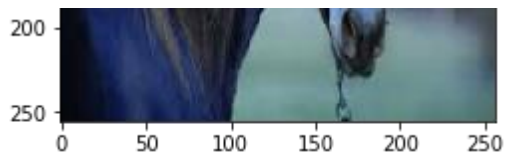


The Predicted Testing image is =humans verify below

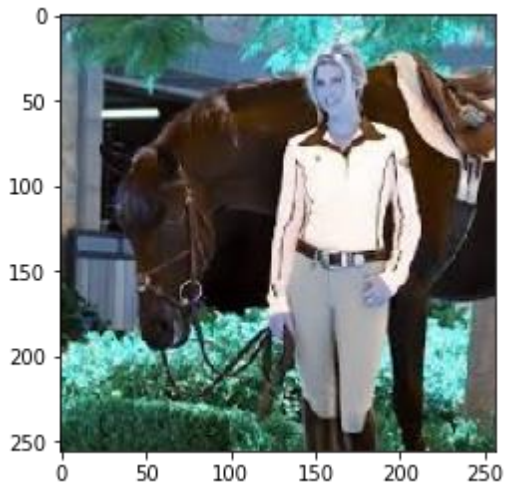


The Predicted Testing image is =humans verify below





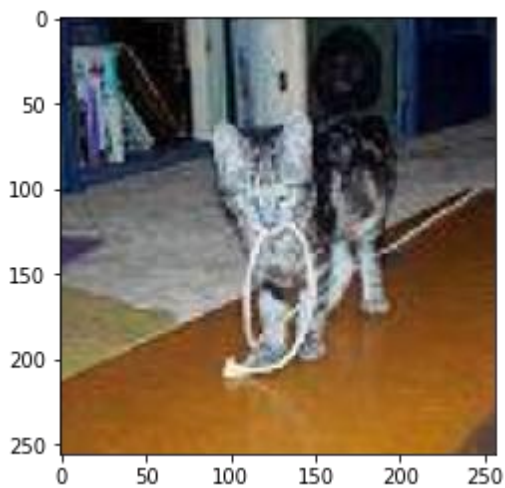
The Predicted Testing image is =humans verify below



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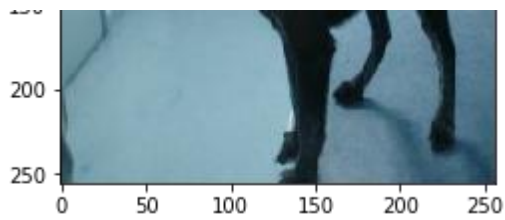


The Predicted Testing image is =cats verify below

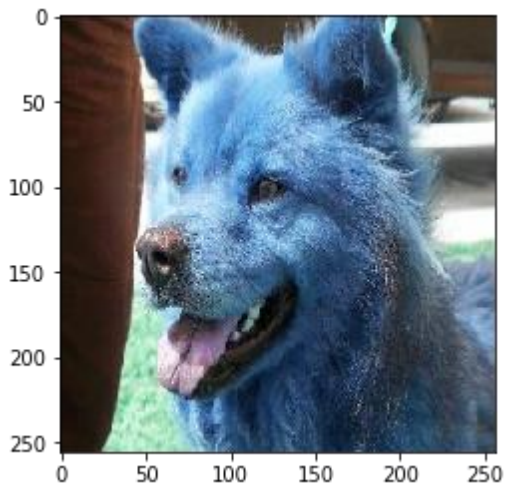


The Predicted Testing image is =humans verify below

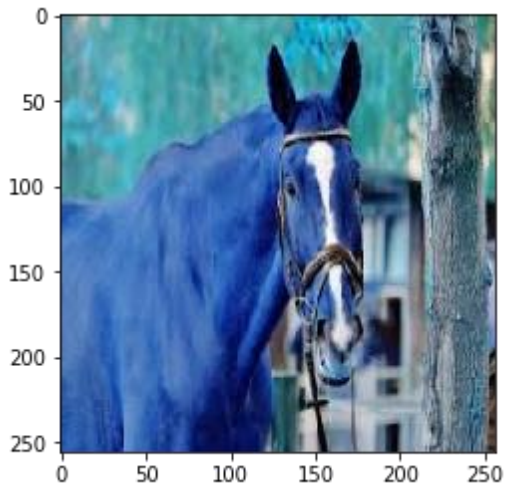




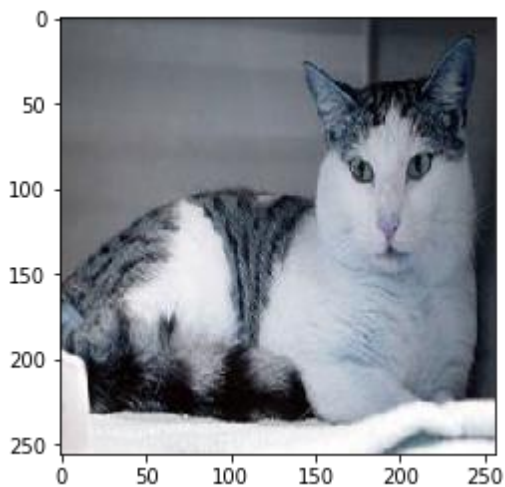
The Predicted Testing image is =humans verify below



The Predicted Testing image is =humans verify below

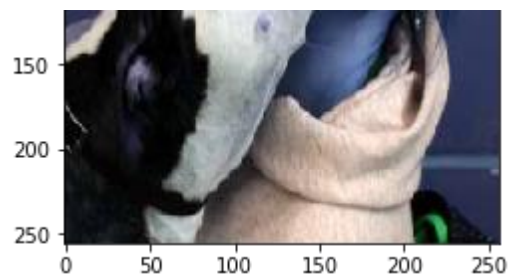


The Predicted Testing image is =humans verify below



The Predicted Testing image is =humans verify below






```

from keras import models
from keras import layers
from keras import optimizers
from keras.applications import VGG16
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from keras.layers import Dense, Conv2D, Flatten, MaxPooling2D, Dropout
from sklearn.utils import shuffle
from PIL import Image
from google.colab.patches import cv2_imshow
import matplotlib.pyplot as plt
import glob
import numpy as np
import os, cv2
#Load the VGG model

def gen_image(img):
    plt.imshow(img)
    return plt

from google.colab import drive
drive.mount('/content/drive')
train_dir = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/train'
validation_dir = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/validatio

img_width, img_height = 256, 256
image_size = img_height

vgg_conv = VGG16(weights='imagenet', include_top=False, input_shape=(img_width, img_height

# Freeze the layers except the last 4 layers
for layer in vgg_conv.layers[:-2]:
    layer.trainable = False

# Check the trainable status of the individual layers
for layer in vgg_conv.layers:
    print(layer, layer.trainable)

# Create the model
model = models.Sequential()

# Add the vgg convolutional base model
model.add(vgg_conv)

# Add new layers
model.add(Flatten())
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(4, activation='softmax'))

```

```
# Show a summary of the model. Check the number of trainable parameters
model.summary()
```

```
train_datagen = ImageDataGenerator(
    rescale=1./255,
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest')
```

```
validation_datagen = ImageDataGenerator(rescale=1./255)
```

```
# Change the batchsize according to your system RAM
train_batchsize = 50
val_batchsize = 10
```

```
train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(image_size, image_size),
    batch_size=train_batchsize,
    class_mode='categorical')
```

```
validation_generator = validation_datagen.flow_from_directory(
    validation_dir,
    target_size=(image_size, image_size),
    batch_size=val_batchsize,
    class_mode='categorical',
    shuffle=False)
```

```
# Compile the model
model.compile(loss='categorical_crossentropy',
              optimizer=optimizers.RMSprop(lr=1e-4),
              metrics=['acc'])
```

```
# Train the model
history = model.fit_generator(
    train_generator,
    steps_per_epoch=train_generator.samples/train_generator.batch_size,
    epochs=5,
    validation_data=validation_generator,
    validation_steps=validation_generator.samples/validation_generator.batch_size,
    verbose=1)
```

```
acc = history.history['acc']
val_acc = history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']
```

```
epochs = range(len(acc))
```

```
plt.plot(epochs, acc, 'b', label='Training acc')
```

```
plt.plot(epochs, val_acc, 'r', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()

plt.figure()

plt.plot(epochs, loss, 'b', label='Training loss')
plt.plot(epochs, val_loss, 'r', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()

plt.show()
```

```
#####333
# insert the testing code
#TESTING
data_path = '/content/drive/My Drive/Colab Notebooks/cats_dogs_horse_humans/test/all_data'
img_data_list=[]
img_list = glob(data_path+'/*.jpg')
for img in img_list:
    input_img=cv2.imread(img,1 )
    input_img_resize=cv2.resize(input_img,(256,256))
    img_data_list.append(input_img_resize)

img_data = np.array(img_data_list)
img_data = img_data.astype('float32')
x_test = shuffle(img_data, random_state=2)
x_test=x_test/255
#Nv_test=x_test.shape[0]
x_test = x_test.reshape(40,256,256,3)
label=['humans','horse','dogs','cats']
ytested = model.predict_classes(x_test)
for i in range(40):
    print("The Predicted Testing image is =%s verify below" %label[ytested[i]])
    gen_image(x_test[i]).show() # printing image vs the predicted image below
```

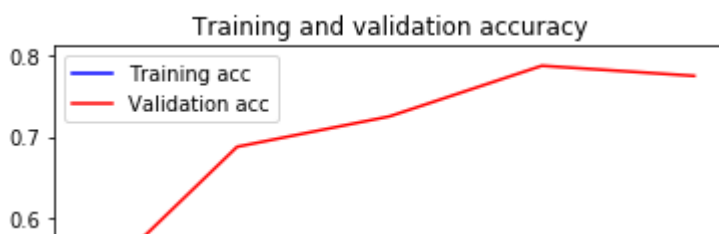


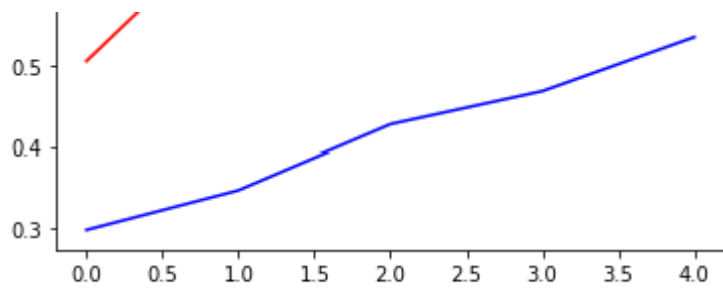
Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.m
 <keras.engine.input_layer.InputLayer object at 0x7ff970bb54e0> False
 <keras.layers.convolutional.Conv2D object at 0x7ff970bb54a8> False
 <keras.layers.convolutional.Conv2D object at 0x7ff970bb5780> False
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 Model: "sequential_3"

Layer (type)	Output Shape	Param #
vgg16 (Model)	(None, 8, 8, 512)	14714688
flatten_3 (Flatten)	(None, 32768)	0
dense_7 (Dense)	(None, 64)	2097216
dropout_5 (Dropout)	(None, 64)	0
dense_8 (Dense)	(None, 64)	4160
dropout_6 (Dropout)	(None, 64)	0
dense_9 (Dense)	(None, 4)	260

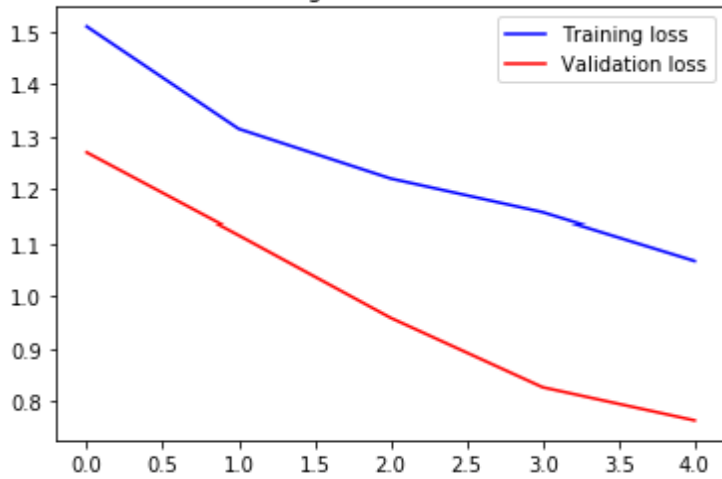
Total params: 16,816,324
 Trainable params: 4,461,444
 Non-trainable params: 12,354,880

Found 608 images belonging to 4 classes.
 Found 160 images belonging to 4 classes.
 Epoch 1/5
 13/12 [=====] - 12s 923ms/step - loss: 1.5013 - acc: 0.287
 Epoch 2/5
 13/12 [=====] - 10s 794ms/step - loss: 1.3094 - acc: 0.356
 Epoch 3/5
 13/12 [=====] - 11s 815ms/step - loss: 1.2177 - acc: 0.432
 Epoch 4/5
 13/12 [=====] - 10s 804ms/step - loss: 1.1500 - acc: 0.470
 Epoch 5/5
 13/12 [=====] - 11s 811ms/step - loss: 1.0769 - acc: 0.534

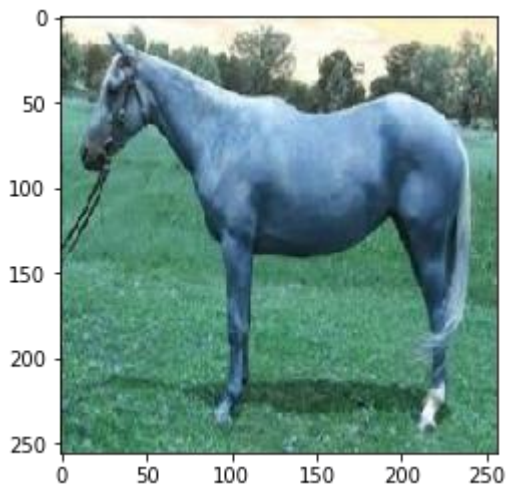




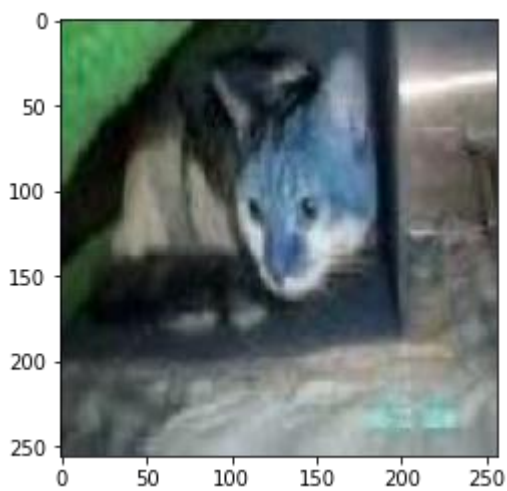
Training and validation loss



The Predicted Testing image is =cats verify below



The Predicted Testing image is =horse verify below

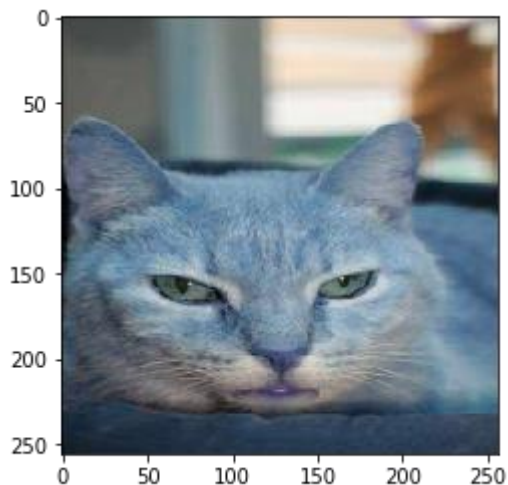


The Predicted Testing image is =dogs verify below

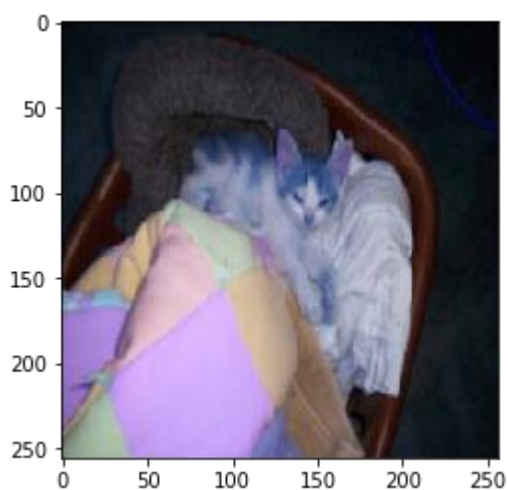




The Predicted Testing image is =horse verify below



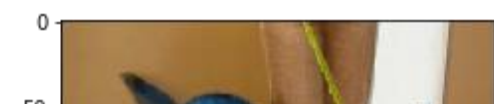
The Predicted Testing image is =horse verify below

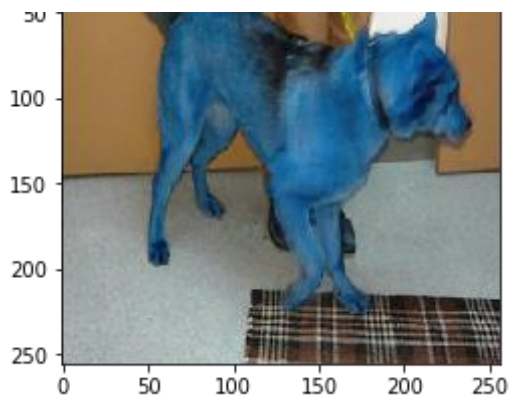


The Predicted Testing image is =cats verify below

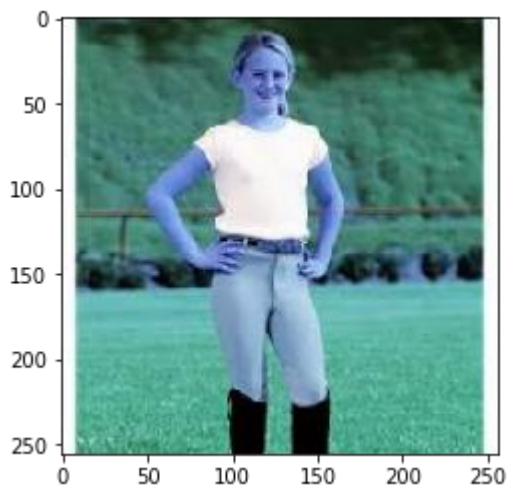


The Predicted Testing image is =dogs verify below

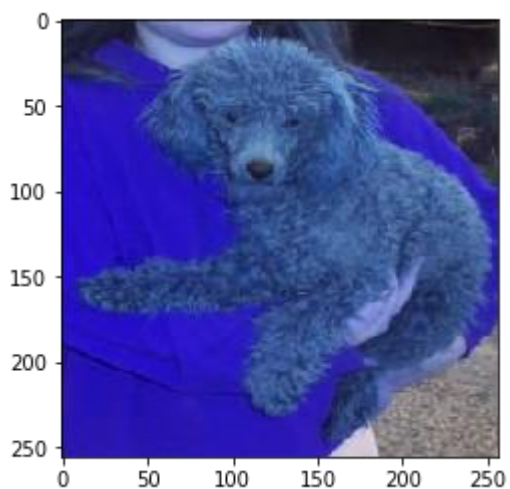




The Predicted Testing image is =humans verify below



The Predicted Testing image is =horse verify below

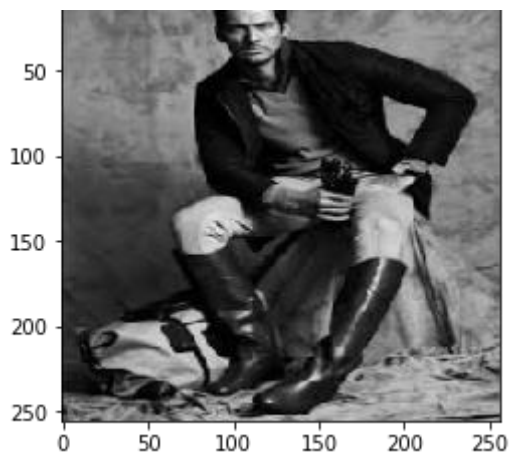


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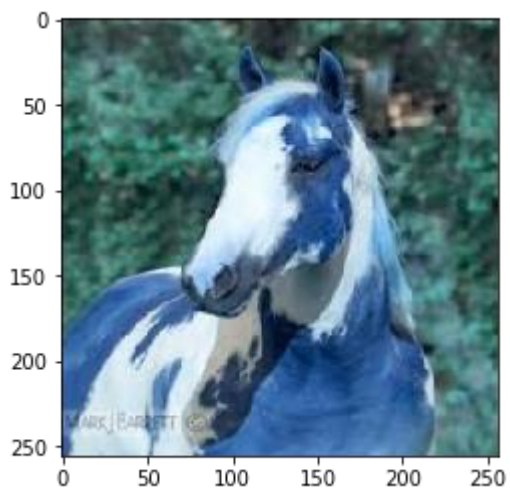


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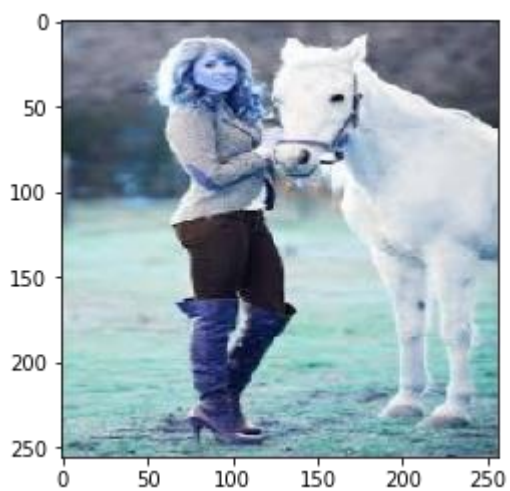




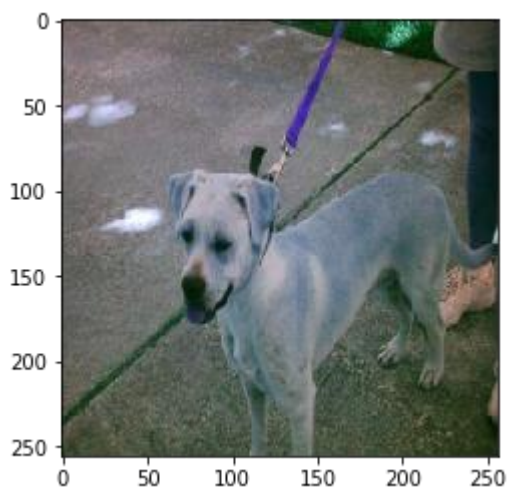
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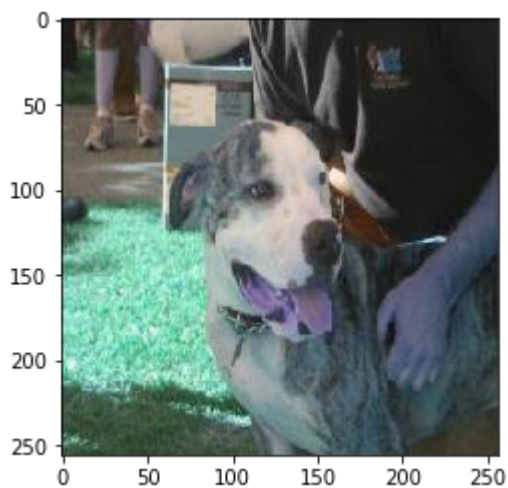
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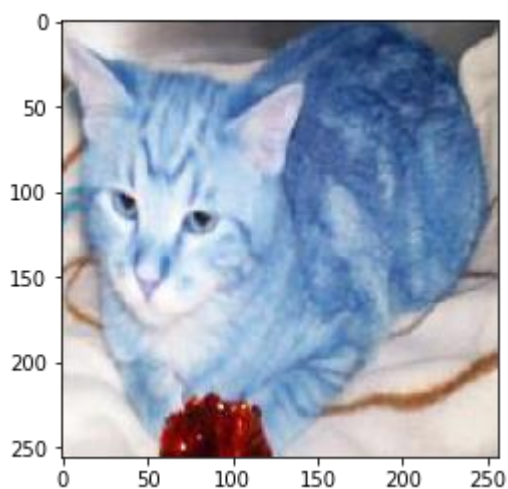
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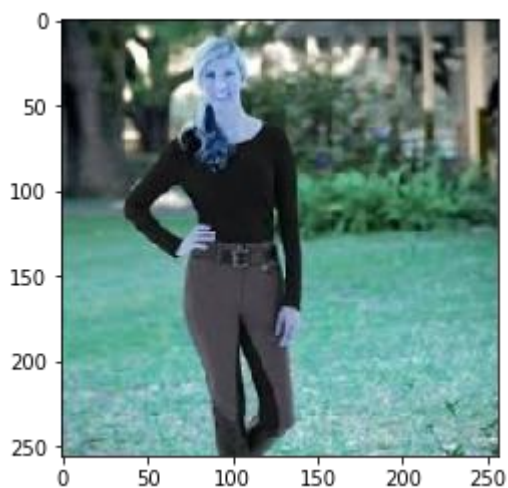
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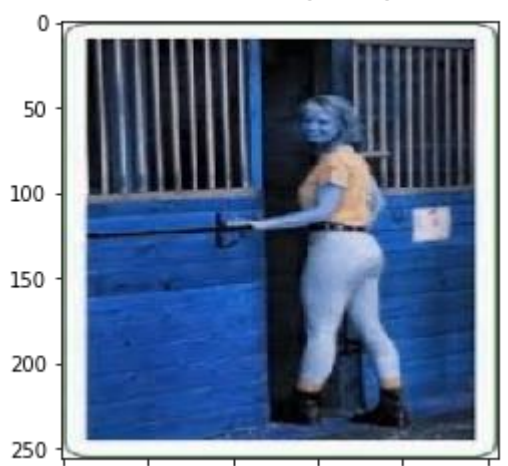
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The Predicted Testing image is =humans verify below

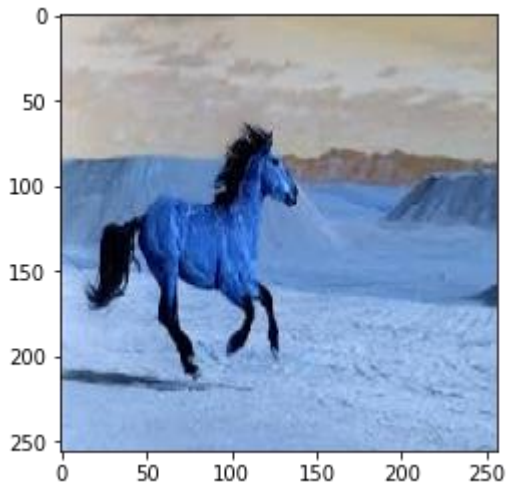


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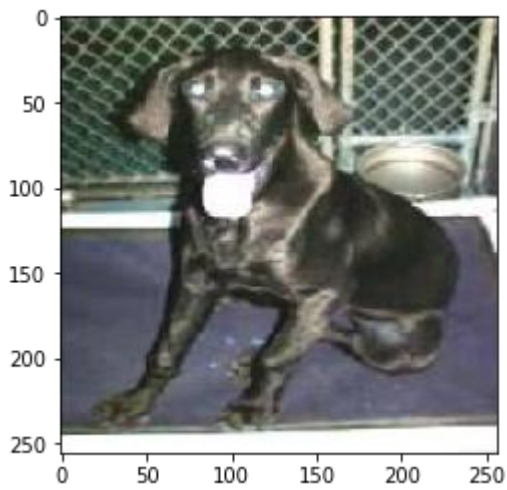


0 50 100 150 200 250

The Predicted Testing image is =cats verify below



The Predicted Testing image is =horse verify below

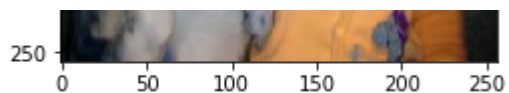


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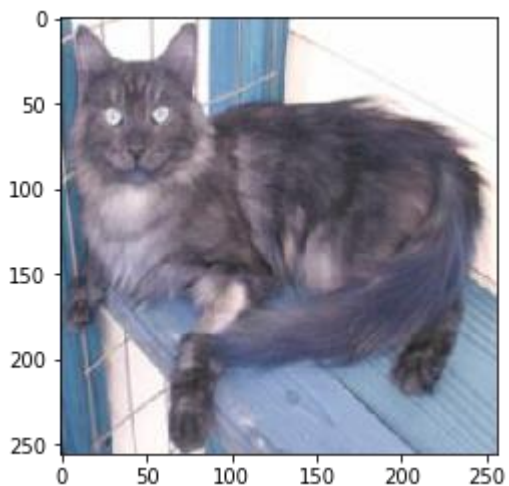


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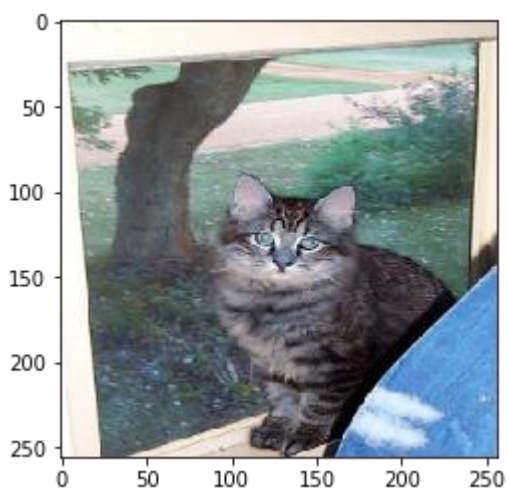




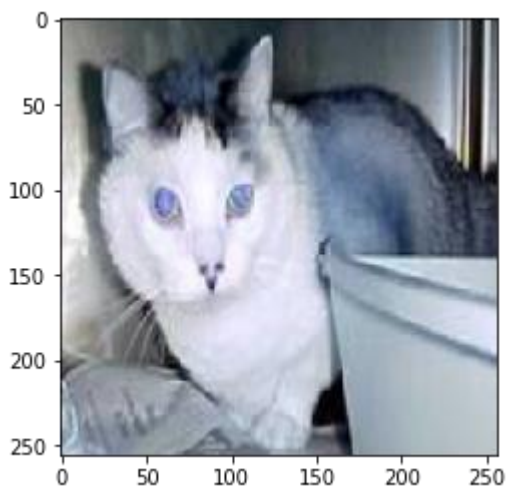
The Predicted Testing image is =horse verify below



The Predicted Testing image is =horse verify below



The Predicted Testing image is =horse verify below

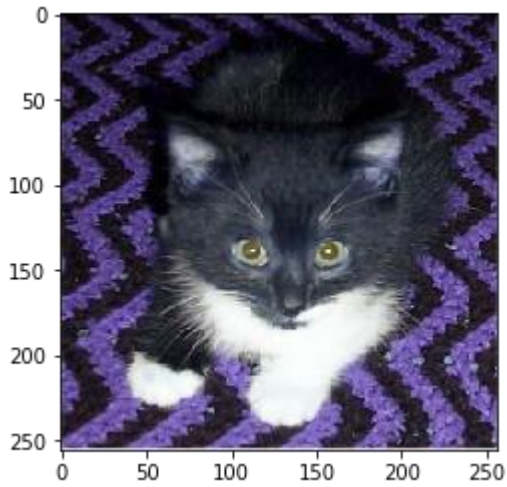


The Predicted Testing image is =humans verify below

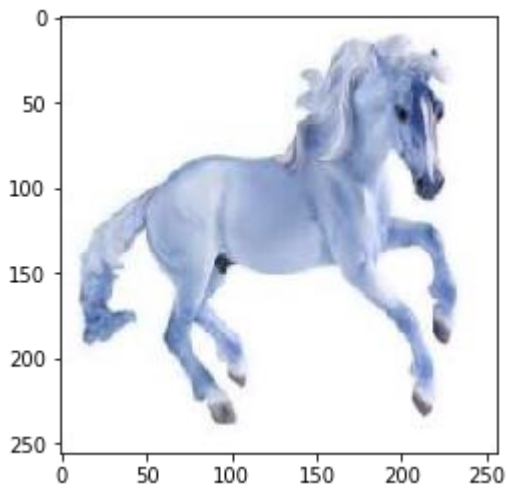




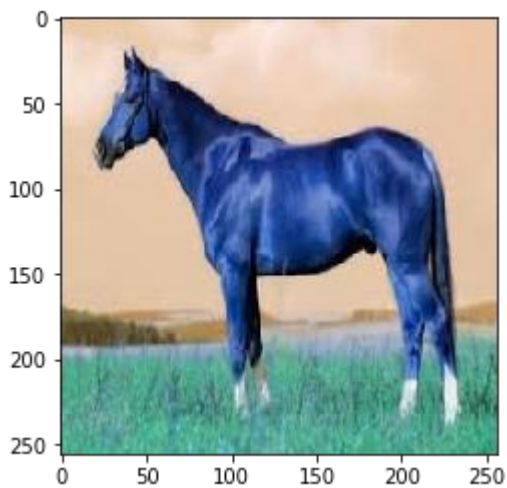
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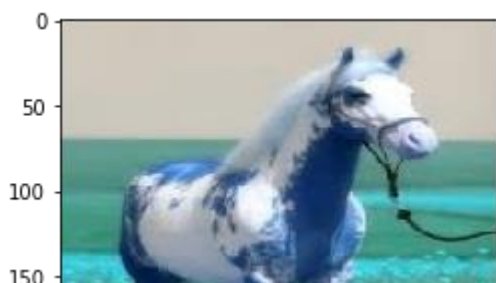
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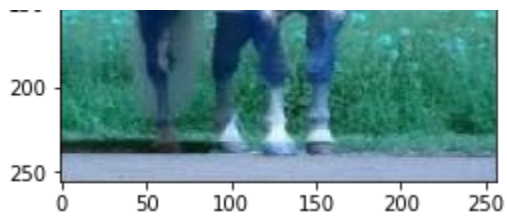


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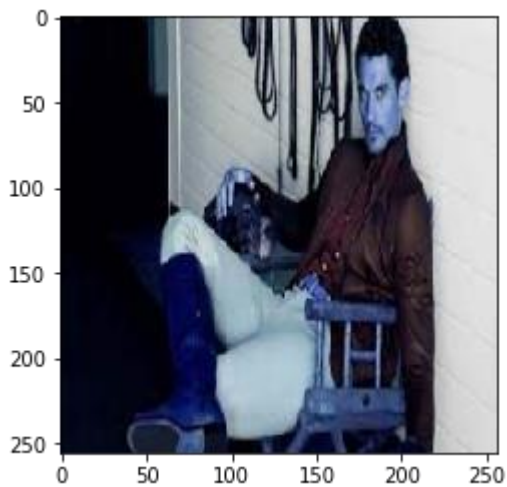


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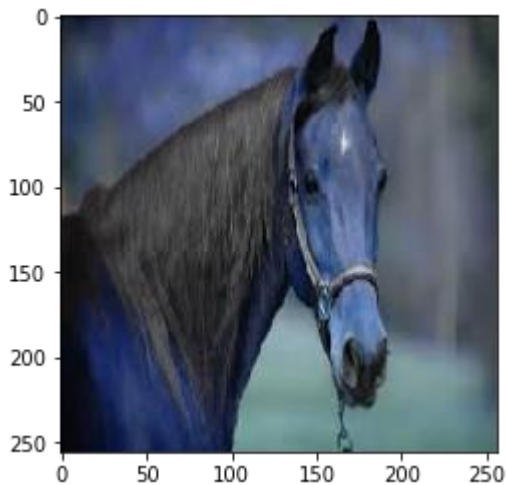




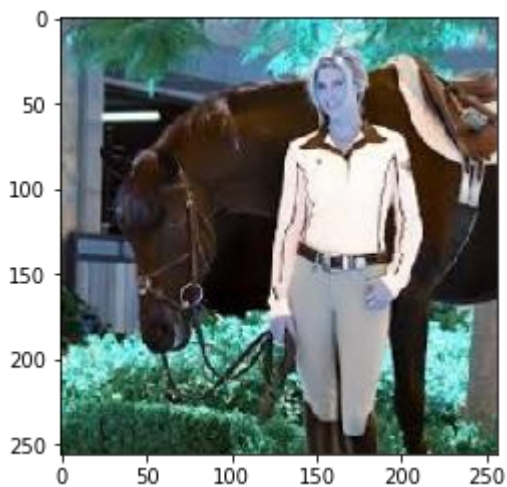
The Predicted Testing image is =humans verify below



The Predicted Testing image is =horse verify below



The Predicted Testing image is =humans verify below

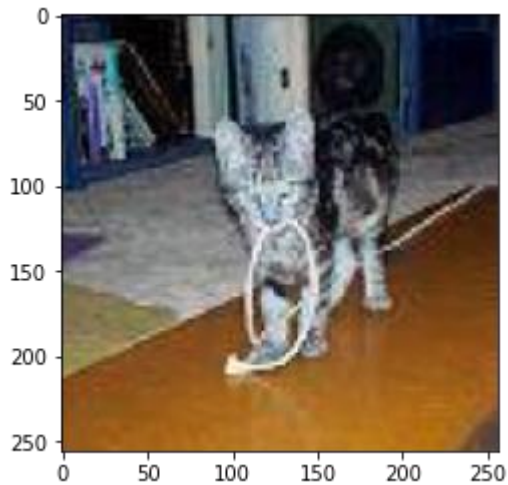


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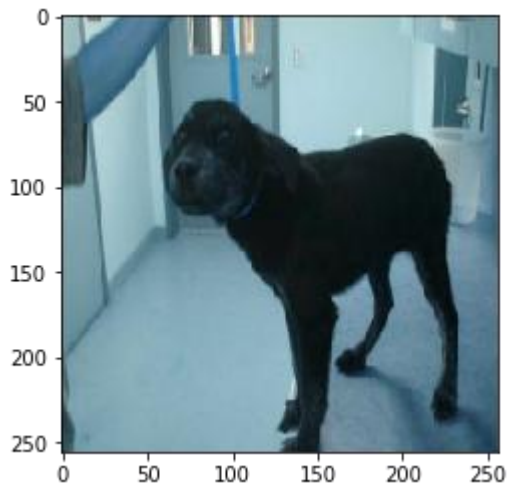




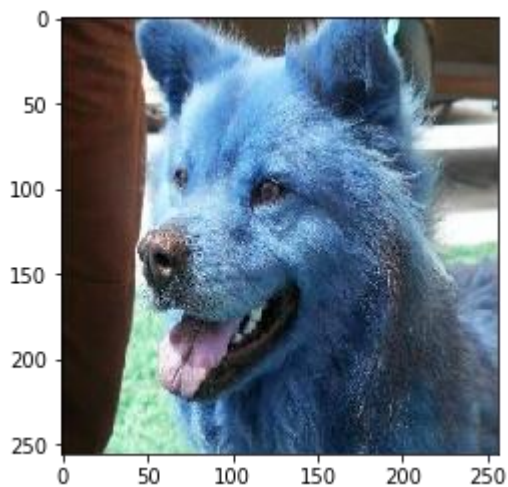
The Predicted Testing image is =horse verify below



The Predicted Testing image is =horse verify below



The Predicted Testing image is =horse verify below

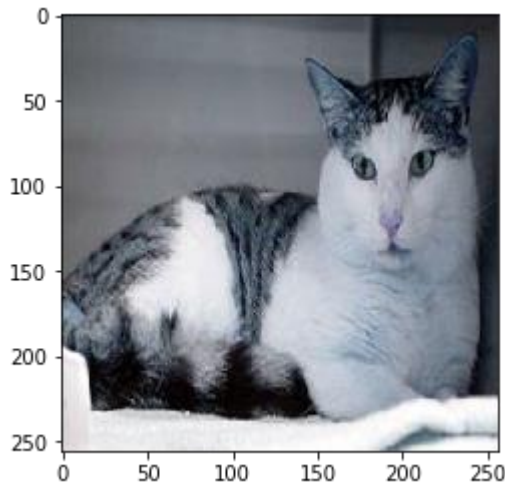


The Predicted Testing image is =horse verify below

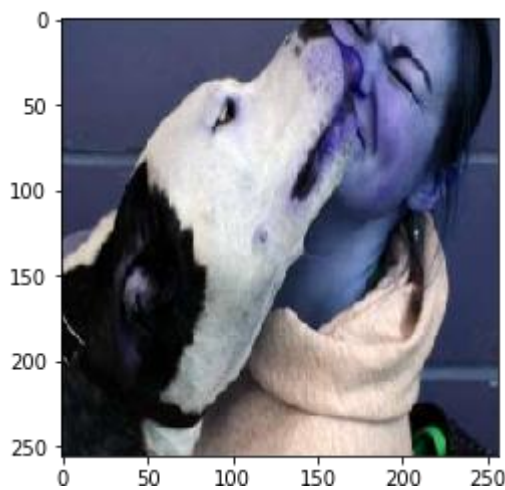




The Predicted Testing image is =horse verify below



The Predicted Testing image is =horse verify below



EXPLANATION:

TASK 1:

- At the beginning we are including few libraries and lines of code to generate reproduceable result, (i.e) to get same accuracy no matter how many times we run the code.
- The `environ['PYTHONHASHSEED']` is used to maintain the precision which helps us to get same results.
- This seed is set by means of numpy a library which generates random numbers at python level.
- We use the tensor flow for all the functions in this program.
- We import all the files necessary for the code to run.
- We give separate data paths for training, testing and validation.
- Now add the model layers to create the CNN, this time we are using 4 convolution layers, 1 pooling layer, 1 flatten, 2 Dropout and 3 dense layers.
- With train and validation batch size we generate the train and validation data.
- In previous assignments we split the training data into training and validation, but in this assignment, we use separate data.
- Then we compile and train the model with the necessary arguments.
- Then we test the data by creating a data list, reshaping it, normalizing it and predicting the results with the image.

TASK 2:

- For task 2 we include all the libraries and the tensor flow functions along with some files which is imported necessary for the code.
- In this task for transfer learning we use VGG 16 model layer, so we need to include all the libraries for the VGG 16 model.
- We then add the model layers and we also include some code for freezing the layer as long it will be false once it's true the flow of code will be normal.
- Then we add the code to analyses the data and represent it in the form of graph.
- We give separate data paths for training, testing and validation.
- Again, with train and validation batch size we generate the train and validation data.
- Then we test the data by creating a data list, reshaping it, normalizing it and predicting the results with the image.

CONCLUSION:

In the first task we are training and validating the data by generating the weights and trying to predict the data. In previous assignments we are training and validating by splitting the data, but in this assignment, we are giving separate directory for these data. When we are testing and printing the result for task 1 we notice that the accuracy is very poor. This is because we are generating the weights. To overcome this in task 2 we are using VGG 16 model layer, which has in built layers that can generate pre trained weights. In task 2 when we test the data, we get a better accuracy than task 1 but not very good accuracy. In this assignment the images are not sectioned into different folders, they are in a single folder. There is a possibility for mismatch. So, we use VGG 16 for better performance.