```
FIRST_NAME = "Mel"

LAST_NAME = "Gerst"

STUDENT_ID = "800995291"

Final Class Project - ITCS 5154

Dog Breed Classifier - Student Mel Gerst

Duplicating project originally by TechVidvan
```

```
# Dog Breed Classifier
# ITCS 5154 - Student Mel Gerst
# Duplicating project by TechVidvan
# Import necessary packages for dog breed classifier
import cv2
import numpy as np
import pandas as pd
import tensorflow
import pathlib
import os
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.models import load_model,Model
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.layers import Dense,GlobalAveragePooling2D,Dropout,BatchNorma
from tensorflow.keras.applications.resnet v2 import ResNet50V2,preprocess input
print("Imports Complete")
print(pathlib.Path().resolve())
from google.colab import drive
drive.mount('/content/drive')
# Assuming your file is in "My Drive/data.csv"
# file path = '/content/drive/My Drive/data.csv'
```

Imports Complete /content

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

```
#specify number
num_breeds = 60
im_size = 224
batch_size = 64
cnceder = LabelEnceder()
```

```
#read the csv file
df_labels = pd.read_csv("/content/drive/My Drive/ColabNotebooks/DogBreedClassifie
#store training and testing images folder location
train_file = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/AI
test_file = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/AI
#check the total number of unique breed in our dataset file
print("Total number of unique Dog Breeds in data:",len(df_labels.breed.unique()))
print(os.listdir(test_file))
```

Start coding or generate with AI.

```
#get only 60 unique breeds record
breed_dict = list(df_labels['breed'].value_counts().keys())
new_list = sorted(breed_dict,reverse=True)[:num_breeds*2+1:2]
#change the dataset to have only those 60 unique breed records
df_labels = df_labels.query('breed in @new_list')
#create new column which will contain image name with the image extension
df_labels['img_file'] = df_labels['id'].apply(lambda x: x + ".jpg")
print("Total number of unique Dog Breeds used in model training:",len(df_labels.b
print("The breeds used for training and testing are:", sorted(df_labels.breed.unice)
```

```
#create a numpy array of the shape
#(number of dataset records, image size , image size, 3 for rgb channel layer)
#this will be input for model
train_x = np.zeros((len(df_labels), im_size, im_size, 3), dtype='float32')
```

Total number of unique Dog Breeds used in model training: 60
The breeds used for training and testing are: ['afghan\_hound', 'airedale', 'a

```
#iterate over img_file column of our dataset
for i, img_id in enumerate(df_labels['img_file']):
    #read the image file and convert into numeric format
    #resize all images to one dimension i.e. 224x224
    #we will get array with the shape of
    #(224,224,3) where 3 is the RGB channels layers
    img = cv2.resize(cv2.imread(train_file+img_id,cv2.IMREAD_COLOR),((im_size,im_size))
    #scale array into the range of -1 to 1.
    #preprocess the array and expand its dimension on the axis 0
    img_array = preprocess_input(np.expand_dims(np.array(img[...,::-1].astype(np.fle))
#update the train_x variable with new element
train_x[i] = img_array
```

```
#building the model using ResNet50V2 with input shape of our image array
#weights for our network will be from of imagenet dataset
#we will not include the first Dense layer
resnet = ResNet50V2(input shape = [im size,im size,3], weights='imagenet', include
#freeze all trainable layers and train only top layers
for layer in resnet.layers:
    layer.trainable = False
#add global average pooling layer and Batch Normalization layer
x = resnet.output
x = BatchNormalization()(x)
x = GlobalAveragePooling2D()(x)
x = Dropout(0.5)(x)
#add fully connected layer
x = Dense(1024, activation='relu')(x)
x = Dropout(0.5)(x)
#add output layer having the shape equal to number of breeds
predictions = Dense(num_breeds, activation='softmax')(x)
```

#create model class with inputs and outputs

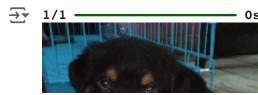
```
model = Model(inputs=resnet.input, outputs=predictions)
#model.summary()
#epochs for model training and learning rate for optimizer
epochs = 10
learning_rate = 1e-3
#using RMSprop optimizer compile or build the model
optimizer = RMSprop(learning rate=learning rate,rho=0.9)
model.compile(optimizer=optimizer,
              loss='sparse_categorical_crossentropy',
              metrics=["accuracy"])
#fit the training generator data and train the model
model.fit(train_generator,
                 steps_per_epoch= x_train.shape[0] // batch_size,
                 epochs= epochs,
                 validation_data= test_generator,
                 validation_steps= x_test.shape[0] // batch_size)
#Save the model for prediction
model.save("model.keras")
#load the model
model = load_model("model.keras")
\rightarrow Epoch 1/10
    /usr/local/lib/python3.10/dist-packages/keras/src/trainers/data_adapters/py_data_
      self._warn_if_super_not_called()
    64/64 -
                             — 1029s 16s/step - accuracy: 0.3116 - loss: 2.9106 -
    Epoch 2/10
     1/64
```

```
11:56 11s/step - accuracy: 0.6250 - loss: 1.4867/us
  self.gen.throw(typ, value, traceback)
                        --- 15s 54ms/step - accuracy: 0.6250 - loss: 1.4867 - \
64/64 -
Epoch 3/10
64/64 -
                          - 1034s 16s/step - accuracy: 0.6197 - loss: 1.2991 -
Epoch 4/10
64/64 -
                          - 14s 44ms/step - accuracy: 0.5938 - loss: 1.1805 - v
Epoch 5/10
64/64 -
                          - 1022s 16s/step - accuracy: 0.6765 - loss: 1.0837 -
Epoch 6/10
                          - 14s 26ms/step - accuracy: 0.6094 - loss: 1.0612 - \
64/64 -
Epoch 7/10
64/64 -
                          - 1015s 16s/step - accuracy: 0.7038 - loss: 0.9997 -
Epoch 8/10
64/64 -
                          - 14s 26ms/step - accuracy: 0.7031 - loss: 1.1159 - \
Epoch 9/10
64/64 -
                          - 1020s 16s/step - accuracy: 0.6974 - loss: 0.9663 -
Epoch 10/10
61/61
                          - 1/e 26ms/stan - accuracy: 0 7021 - loss: 0 7606 - v
```

**το Συπογοία – αυταίατο, υπίσος – τουσο, υπίσος – τουσο, υπίσου –** 

```
#get the image of the dog #1 for prediction
pred_img_path = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/#
#read the image file and convert into numeric format
#resize all images to one dimension i.e. 224x224
pred_img_array = cv2.resize(cv2.imread(pred_img_path,cv2.IMREAD_COLOR),((im_size,in
#scale array into the range of -1 to 1.
#expand the dimesion on the axis 0 and normalize the array values
pred_img_array = preprocess_input(np.expand_dims(np.array(pred_img_array[...,::-1].
#feed the model with the image array for prediction
pred_val = model.predict(np.array(pred_img_array,dtype="float32"))
#display the image of dog
from google.colab.patches import cv2 imshow
cv2 imshow(cv2.resize(cv2.imread(pred img path,cv2.IMREAD COLOR),((im size,im size)
# cv2.imshow("TechVidvan",cv2.resize(cv2.imread(pred_img_path,cv2.IMREAD_COLOR),((i))
#display the predicted breed of dog
pred breed = sorted(new list)[np.argmax(pred val)]
print("Predicted Breed for this Dog is :",pred breed)
#get the image of the dog #2 for prediction
pred_img_path2 = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/
#read the image file and convert into numeric format
#resize all images to one dimension i.e. 224x224
pred_img_array2 = cv2.resize(cv2.imread(pred_img_path2,cv2.IMREAD_COLOR),((im_size,
#scale array into the range of -1 to 1.
#expand the dimesion on the axis 0 and normalize the array values
pred_img_array2 = preprocess_input(np.expand_dims(np.array(pred_img_array2[...,::-1
#feed the model with the image array for prediction
pred val2 = model.predict(np.array(pred img array2,dtype="float32"))
#display the image of dog
from google.colab.patches import cv2_imshow
cv2_imshow(cv2.resize(cv2.imread(pred_img_path2,cv2.IMREAD_COLOR),((im_size,im_size
# cv2.imshow("TechVidvan",cv2.resize(cv2.imread(pred_img_path,cv2.IMREAD_COLOR),((i))
#display the predicted breed of dog
pred_breed2 = sorted(new_list)[np.argmax(pred_val2)]
print("Predicted Breed for this Dog is :",pred_breed2)
#get the image of the dog #3 for prediction
pred_img_path3 = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/
#read the image file and convert into numeric format
#resize all images to one dimension i.e. 224x224
pred_img_array3 = cv2.resize(cv2.imread(pred_img_path3,cv2.IMREAD_COLOR),((im_size,
#scale array into the range of -1 to 1.
#expand the dimesion on the axis 0 and normalize the array values
pred_img_array3 = preprocess_input(np.expand_dims(np.array(pred_img_array3[...,::-1
#feed the model with the image array for prediction
```

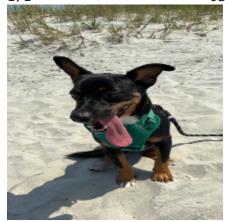
```
pred_val3 = model.predict(np.array(pred_img_array3,dtype="float32"))
#display the image of dog
from google.colab.patches import cv2_imshow
cv2_imshow(cv2.resize(cv2.imread(pred_img_path3,cv2.IMREAD_COLOR),((im_size,im_size
# cv2.imshow("TechVidvan",cv2.resize(cv2.imread(pred_img_path,cv2.IMREAD_COLOR),((i))
#display the predicted breed of dog
pred_breed3 = sorted(new_list)[np.argmax(pred_val3)]
print("Predicted Breed for this Dog is :",pred_breed3)
#get the image of the dog #4 for prediction
pred_img_path4 = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/
#read the image file and convert into numeric format
#resize all images to one dimension i.e. 224x224
pred img array4 = cv2.resize(cv2.imread(pred img path4,cv2.IMREAD COLOR),((im size,
#scale array into the range of -1 to 1.
#expand the dimesion on the axis 0 and normalize the array values
pred_img_array4 = preprocess_input(np.expand_dims(np.array(pred_img_array4[...,::-1
#feed the model with the image array for prediction
pred_val4 = model.predict(np.array(pred_img_array4,dtype="float32"))
#display the image of dog
from google.colab.patches import cv2 imshow
cv2_imshow(cv2.resize(cv2.imread(pred_img_path4,cv2.IMREAD_COLOR),((im_size,im_size
# cv2.imshow("TechVidvan",cv2.resize(cv2.imread(pred_img_path,cv2.IMREAD_COLOR),((i))
#display the predicted breed of dog
pred breed4 = sorted(new list)[np.argmax(pred val4)]
print("Predicted Breed for this Dog is :",pred_breed4)
#get the image of the dog #5 for prediction
pred_img_path5 = '/content/drive/My Drive/ColabNotebooks/DogBreedClassifierProject/
#read the image file and convert into numeric format
#resize all images to one dimension i.e. 224x224
pred_img_array5 = cv2.resize(cv2.imread(pred_img_path5,cv2.IMREAD_COLOR),((im_size,
#scale array into the range of -1 to 1.
#expand the dimesion on the axis 0 and normalize the array values
pred_img_array5 = preprocess_input(np.expand_dims(np.array(pred_img_array5[...,::-1
#feed the model with the image array for prediction
pred_val5 = model.predict(np.array(pred_img_array5,dtype="float32"))
#display the image of dog
from google.colab.patches import cv2 imshow
cv2_imshow(cv2.resize(cv2.imread(pred_img_path5,cv2.IMREAD_COLOR),((im_size,im_size)
#display the predicted breed of dog
pred_breed5 = sorted(new_list)[np.argmax(pred_val5)]
print("Predicted Breed for this Dog is :",pred_breed5)
```



Os 198ms/step



Predicted Breed for this Dog is : rottweiler
1/1 \_\_\_\_\_\_ 0s 206ms/step



Predicted Breed for this Dog is : miniature\_pinscher
1/1 \_\_\_\_\_\_ 0s 190ms/step



Predicted Breed for this Dog is : labrador\_retriever
1/1 \_\_\_\_\_\_ 0s 177ms/step



Predicted Breed for this Dog is : scottish\_deerhound

1/1 \_\_\_\_\_\_ 0s 321ms/step



Predicted Breed for this Dog is : whippet