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In [1]: #Importing Libraries

import pandas as pd
import numpy as np
import cv2
import tensorflow as tf
import keras
import h5py
from keras.applications.vgg16 import VGG16, preprocess_input
from keras.models import Sequential, Model
from keras.layers import Conv2D, Flatten, MaxPool2D, Input, Dropout, Dense
from keras.layers.normalization import BatchNormalization
from keras.callbacks import EarlyStopping, Callback, ModelCheckpoint
from sklearn.preprocessing import OneHotEncoder
from sklearn.preprocessing import LabelEncoder
from keras.utils import to_categorical
from keras.optimizers import Adam
import os
from pathlib import Path
import glob
from sklearn.model_selection import train_test_split
import cv2
import matplotlib.pyplot as plt
import time
os.environ['TF_CPP_MIN_LOG_LEVEL'] = '3'

print('Imported Libraries')

print('Reading the data')
#Reading the data
member = ('/kaggle/input/garbage-classification/Garbage classification/Garbage class
catagories = os.listdir(member)
list_items = []
for cat in catagories:
    catagory_img = (member + cat)
    for _ in (glob.glob(catagory_img + '/' + '*.jpg')):
        list_items.append([cat, _])

#Convert list into dataframe
data = pd.DataFrame(list_items, columns = ['catagory', 'filepath'], index = None)
data = data.sample(frac=1).reset_index(drop=True)
data.head(5)
data.shape

#print('Splitting the dataset')
train_data = data[1:2000]
val_data = data[2001:2200]
test_data = data[2201:2527]

```

Using TensorFlow backend.  
Imported Libraries  
Reading the data

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In [2]: #Preprocessing for Training set

def adv_preprocessing(image):
    #loading images with

    preimgs = []
    img = cv2.imread(image, cv2.IMREAD_UNCHANGED)

    #Setting dimensions to resize

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height = 224
width = 224

dim = (width, height)
res = cv2.resize(img, dim, interpolation = cv2.INTER_LINEAR)
preimgs.append(res)

#Removing noise from image - Gaussian blur

blurred_img = cv2.GaussianBlur(res, (5,5),0)
preimgs.append(blurred_img)

#Segmentation
#-----
image = res
gray = cv2.cvtColor(image, cv2.COLOR_RGB2GRAY)
ret,thresh = cv2.threshold(gray, 0,255,cv2.THRESH_BINARY+ cv2.THRESH_OTSU)

#More noise removal
#-----
kernel = np.ones((3,3), np.uint8)
opening = cv2.morphologyEx(thresh, cv2.MORPH_OPEN, kernel, iterations=2)

#Sure background area
sure_bg = cv2.dilate(opening, kernel, iterations = 3)

#Finding foreground area
dist_transform = cv2.distanceTransform(opening, cv2.DIST_L2, 5)
ret, sure_fg = cv2.threshold(dist_transform, 0.7 * dist_transform.max(), 255, 0)

# Finding unknown region
sure_fg = np.uint8(sure_fg)
unknown = cv2.subtract(sure_bg, sure_fg)

#Seperating different objects with different backgrounds
#Markers Labelling
ret, markers = cv2.connectedComponents(sure_fg)
#Add one to all labels so that sure background is 0 not 1
markers = markers+1

#Mark the unknown region with 0
markers[unknown == 255] = 0

markers = cv2.watershed(res, markers)
res[markers == -1] = [255,0,0]
placeholder = np.random.rand(224,224)
#Displaying the markers on image
markers = np.dstack([markers,np.zeros((224,224)), placeholder])
#Adding
preimgs.append(res)
preimgs.append(markers)

return preimgs

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#-----
#Preprocessing for Validation and test data

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def norm_data(data):
    batch_data = np.zeros((len(data),224,224,3))
    batch_labels = np.zeros((len(data),6))
    label_enc = LabelEncoder()
    y = label_enc.fit_transform(data['catagory']) #Converting data into Labels
    y = y.reshape(-1,1)

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onehotenc = OneHotEncoder(handle_unknown= 'ignore')
batch_labels = onehotenc.fit_transform(y).toarray() #Onehot Encoding data

#Image normalization
height = 224
width = 224
dim = (height, width)

for i in range(len(data)):
    img = cv2.imread(data.iloc[i]['filepath'])
    res_img = cv2.resize(img, dim, interpolation = cv2.INTER_LINEAR)
    res_img = res_img.astype(np.float32)/255
    batch_data[i] = res_img

return batch_data, batch_labels

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In [3]: #Generator for training data

def adv_gendata(data, batch_size):
    labelenc = LabelEncoder()
    n = len(data)
    steps = n//batch_size
    data['labels'] = labelenc.fit_transform(data['catagory'])
    enc = OneHotEncoder(handle_unknown='ignore')
    enc_df = pd.DataFrame(enc.fit_transform(data[['labels']]).toarray())

    #Defining numpy array to contain Label and image data

    batch_data = np.zeros((batch_size, 224,224,3), dtype = np.float32)
    batch_labels = np.zeros((batch_size, 6))
    indices = np.arange(n)

    c1 = 0
    i = 0
    #Initialize counter
    while True:
        np.random.shuffle(indices)
        next_batch = indices[(i*batch_size):(i+1)*batch_size]
        count = 0

        for j, idx in enumerate(next_batch):
            if count <= batch_size-4:
                img_name = data.iloc[idx]['filepath']
                aug_img = adv_preprocessing(data['filepath'].iloc[idx])
                encoded_label = enc_df.iloc[idx]
                batch_data[count+0] = aug_img[0]
                batch_labels[count+0] = encoded_label
                batch_data[count+1] = aug_img[1]
                batch_labels[count+1] = encoded_label
                batch_data[count+2] = aug_img[2]
                batch_labels[count+2] = encoded_label
                batch_data[count+3] = aug_img[3]
                batch_labels[count+3] = encoded_label
                count +=4
                c1 = c1+1

            else:
                count+=1

        if count==batch_size:
            i += 1
            break

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        i+=1
        yield batch_data, batch_labels

    if i>=steps:
        i=0

```

In [4]:

```

print('Building the model')

#Working on the model
def build_model():
    model = Sequential()
    input_size = Input(shape = (224,224,3), name = 'Input_Image')

    #Layer 1 - Deapth Layer 1,2
    x = Conv2D(64,(3,3), activation = 'relu', padding = 'same', name = 'ConvLayer1')
    x = Conv2D(64,(3,3), activation = 'relu', padding = 'same', name = 'ConvLayer2')
    x = MaxPool2D((2,2), name = 'Maxpool1')(x)
    x = BatchNormalization(name = 'bn1')(x)

    #Layer 2 - Deapth Layer 3,4
    x = Conv2D(128,(3,3), activation = 'relu', padding = 'same', name = 'ConvLayer3')
    x = Conv2D(128,(3,3), activation = 'relu', padding = 'same', name = 'ConvLayer4')
    x = MaxPool2D((2,2), name = 'Maxpool2')(x)
    x = BatchNormalization(name = 'bn2')(x)
    # Layer 3 - Deapth Layer 3
    x = Conv2D(256,(3,3), activation='relu',padding = 'same', name = 'ConvLayer5')
    x = MaxPool2D((2,2), name = 'Maxpool3')(x)
    x = BatchNormalization(name = 'bn3')(x)

    #Flatten the model

    x = Flatten(name = 'Flatten')(x)
    x = Dense(256, activation = 'relu', name = 'FC1')(x)
    x = Dropout(0.7, name = 'Dropout1')(x)
    x = Dense(256, activation = 'relu', name = 'FC2')(x)
    x = Dropout(0.7, name = 'Dropout2')(x)
    x = Dense(6, activation = 'softmax', name = 'Fc3')(x)

    model = Model(input = input_size , output = x)
    return model

#Building the model and summary

model = build_model()
model.summary()

#Initialize the first layer with the weights of Imagenet

f = h5py.File('/kaggle/input/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5')

#Selecting the layers for which weights needs to be set
w,b = f['block1_conv1']['block1_conv1_W_1:0'], f['block1_conv1']['block1_conv1_b_1:0']
model.layers[1].set_weights = [w,b]

w,b = f['block1_conv2']['block1_conv2_W_1:0'], f['block1_conv2']['block1_conv2_b_1:0']
model.layers[2].set_weights = [w,b]

f.close()
model.summary()

```

```
#Compiling the model
from keras.optimizers import SGD
opt = SGD(lr=0.01)
model.compile(loss = "categorical_crossentropy", optimizer = opt)
es = EarlyStopping(patience=5)
chkpt = ModelCheckpoint(filepath= 'bestmodel',save_best_only=True, save_weights_only
model.compile(loss= 'binary_crossentropy', metrics= ['accuracy'], optimizer= opt)
```

Building the model  
Model: "model\_1"

Layer (type)	Output Shape	Param #
Input_Image (InputLayer)	(None, 224, 224, 3)	0
ConvLayer1 (Conv2D)	(None, 224, 224, 64)	1792
ConvLayer2 (Conv2D)	(None, 224, 224, 64)	36928
Maxpool1 (MaxPooling2D)	(None, 112, 112, 64)	0
bn1 (BatchNormalization)	(None, 112, 112, 64)	256
ConvLayer3 (Conv2D)	(None, 112, 112, 128)	73856
ConvLayer4 (Conv2D)	(None, 112, 112, 128)	147584
Maxpool12 (MaxPooling2D)	(None, 56, 56, 128)	0
bn2 (BatchNormalization)	(None, 56, 56, 128)	512
ConvLayer5 (Conv2D)	(None, 56, 56, 256)	295168
Maxpool3 (MaxPooling2D)	(None, 28, 28, 256)	0
bn3 (BatchNormalization)	(None, 28, 28, 256)	1024
Flatten (Flatten)	(None, 200704)	0
FC1 (Dense)	(None, 256)	51380480
Dropout1 (Dropout)	(None, 256)	0
FC2 (Dense)	(None, 256)	65792
Dropout2 (Dropout)	(None, 256)	0
Fc3 (Dense)	(None, 6)	1542
Total params: 52,004,934		
Trainable params: 52,004,038		
Non-trainable params: 896		

Model: "model\_1"

Layer (type)	Output Shape	Param #
Input_Image (InputLayer)	(None, 224, 224, 3)	0
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ConvLayer2 (Conv2D)	(None, 224, 224, 64)	36928
Maxpool1 (MaxPooling2D)	(None, 112, 112, 64)	0
bn1 (BatchNormalization)	(None, 112, 112, 64)	256
ConvLayer3 (Conv2D)	(None, 112, 112, 128)	73856

ConvLayer4 (Conv2D)	(None, 112, 112, 128)	147584
Maxpool12 (MaxPooling2D)	(None, 56, 56, 128)	0
bn2 (BatchNormalization)	(None, 56, 56, 128)	512
ConvLayer5 (Conv2D)	(None, 56, 56, 256)	295168
Maxpool13 (MaxPooling2D)	(None, 28, 28, 256)	0
bn3 (BatchNormalization)	(None, 28, 28, 256)	1024
Flatten (Flatten)	(None, 200704)	0
FC1 (Dense)	(None, 256)	51380480
Dropout1 (Dropout)	(None, 256)	0
FC2 (Dense)	(None, 256)	65792
Dropout2 (Dropout)	(None, 256)	0
Fc3 (Dense)	(None, 6)	1542
=====		
Total params: 52,004,934		
Trainable params: 52,004,038		
Non-trainable params: 896		

/opt/conda/lib/python3.6/site-packages/ipykernel\_launcher.py:34: UserWarning: Update your `Model` call to the Keras 2 API: `Model(inputs=Tensor("In...", outputs=Tensor("Fc..."))`

```
In [5]: #Training the model

batch_size = 20
train_data_gen = adv_gendata(train_data,20)
val_data, val_labels = norm_data(val_data)
test_data, test_labels = norm_data(test_data)

nb_train_steps = train_data.shape[0]//batch_size

epochs = 10
model.fit_generator(train_data_gen,epochs=epochs, verbose = 1, steps_per_epoch=nb_tr
```

Epoch 1/10

/opt/conda/lib/python3.6/site-packages/ipykernel\_launcher.py:7: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: [http://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
import sys
99/99 [=====] - 747s 8s/step - loss: 2.0365 - accuracy: 0.7
969 - val_loss: 1.8714 - val_accuracy: 0.7487
```

Epoch 2/10

```
99/99 [=====] - 743s 8s/step - loss: 0.9206 - accuracy: 0.9
219 - val_loss: 4.4187 - val_accuracy: 0.7119
```

Epoch 3/10

```
99/99 [=====] - 744s 8s/step - loss: 0.8683 - accuracy: 0.9
311 - val_loss: 4.4187 - val_accuracy: 0.7119
```

Epoch 4/10

```
99/99 [=====] - 741s 7s/step - loss: 0.7981 - accuracy: 0.9
378 - val_loss: 4.4187 - val_accuracy: 0.7119
```

Epoch 5/10

```
99/99 [=====] - 741s 7s/step - loss: 0.7982 - accuracy: 0.9
383 - val_loss: 3.8535 - val_accuracy: 0.7487
```

Epoch 6/10

```
99/99 [=====] - 743s 8s/step - loss: 0.8400 - accuracy: 0.9
369 - val_loss: 4.4187 - val_accuracy: 0.7119
Epoch 7/10
99/99 [=====] - 745s 8s/step - loss: 0.8146 - accuracy: 0.9
407 - val_loss: 4.4187 - val_accuracy: 0.7119
Epoch 8/10
99/99 [=====] - 746s 8s/step - loss: 0.7704 - accuracy: 0.9
436 - val_loss: 4.4187 - val_accuracy: 0.7119
Epoch 9/10
99/99 [=====] - 744s 8s/step - loss: 0.7882 - accuracy: 0.9
426 - val_loss: 4.4187 - val_accuracy: 0.7119
Epoch 10/10
99/99 [=====] - 744s 8s/step - loss: 0.8022 - accuracy: 0.9
412 - val_loss: 4.4187 - val_accuracy: 0.7119
```

Out[5]: <keras.callbacks.callbacks.History at 0x7f6f6c3a3eb8>

In [6]: *#Evaluating the model on test data*

```
score = model.evaluate(test_data, test_labels, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

Test loss: 4.312577487500898

Test accuracy: 0.7188138961791992