## waste-classifiction

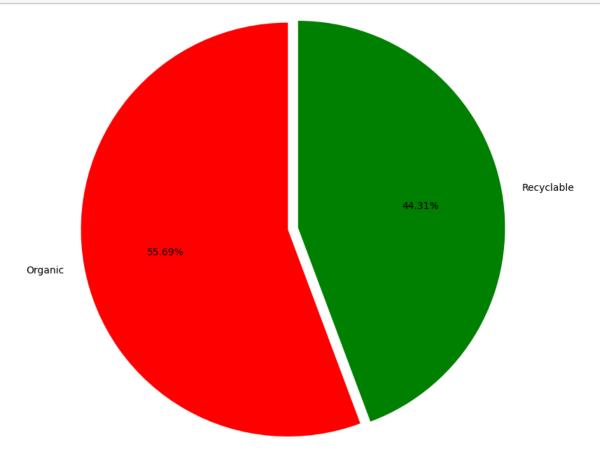
## November 17, 2024

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     from tqdm import tqdm
     import cv2
     import warnings
     warnings.filterwarnings('ignore')
     import os
     for dirname, _, _ in os.walk('/kaggle/input'):
             print(dirname)
[2]: train_path = r"C:\Users\Divish\Desktop\ml project\DATASET\TRAIN"
     test path = r"C:\Users\Divish\Desktop\ml project\DATASET\TEST"
[3]: from keras.models import Sequential
     from keras.layers import Conv2D, MaxPooling2D, Activation, Dropout, Flatten,
     →Dense, BatchNormalization
     from tensorflow.keras.preprocessing.image import ImageDataGenerator
     from tensorflow.keras.utils import plot_model
     from glob import glob
[4]: x_data = []
     y_{data} = []
     for category in glob(train_path+'/*'):
         for file in tqdm(glob(category+'/*')):
             img_array=cv2.imread(file)
             img_array = cv2.cvtColor(img_array, cv2.COLOR_BGR2RGB)
             x_data.append(img_array)
             y_data.append(category.split("/")[-1])
     data=pd.DataFrame({'image': x_data,'label': y_data})
    100%|
    | 12565/12565 [00:23<00:00, 541.33it/s]
    100%|
     | 9999/9999 [00:17<00:00, 585.84it/s]
```

```
[5]: data.shape
```

**[5]**: (22564, 2)

```
[6]: from collections import Counter Counter(y_data)
```



```
[8]: plt.figure(figsize=(20,15))
for i in range(10):
    plt.subplot(4,3,(i%12)+1)
    index=np.random.randint(15000)
```

```
This image is of CitizensDivisitiDestagried projectCMASETITIAMNO
This image is of CitizensDivisi
```

```
[9]: className = glob(train_path + '/*' )
numberOfClass = len(className)
print("Number Of Class: ",numberOfClass)
```

Number Of Class: 2

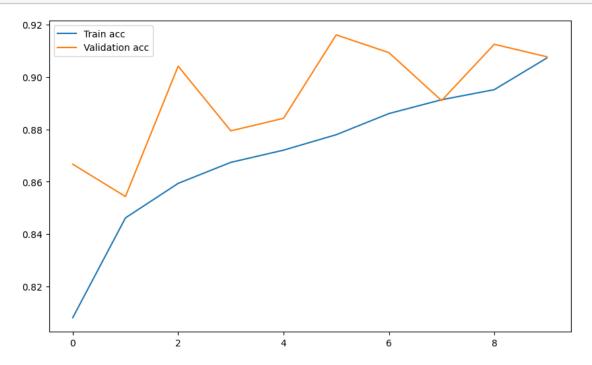
```
[15]: model = Sequential()
  model.add(Conv2D(32,(3,3),input_shape = (128,128,3)))
  model.add(Activation("relu"))
  model.add(MaxPooling2D())

model.add(Conv2D(64,(3,3)))
  model.add(Activation("relu"))
  model.add(MaxPooling2D())
```

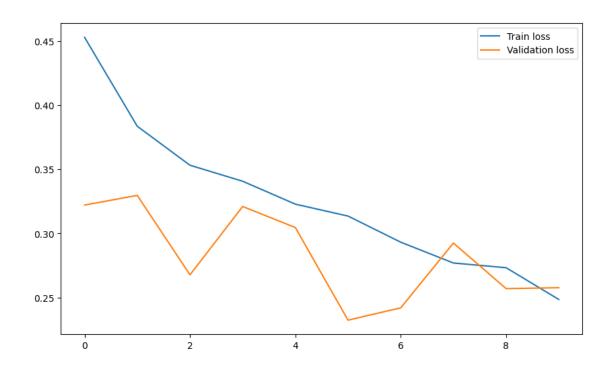
```
model.add(Conv2D(128,(3,3)))
      model.add(Activation("relu"))
      model.add(MaxPooling2D())
     model.add(Flatten())
      model.add(Dense(256))
      model.add(Activation("relu"))
     model.add(Dropout(0.5))
      model.add(Dense(64))
      model.add(Activation("relu"))
      model.add(Dropout(0.5))
      model.add(Dense(numberOfClass)) # output
      model.add(Activation("sigmoid"))
      model.compile(loss = "binary_crossentropy",
                    optimizer = "adam",
                    metrics = ["accuracy"])
[16]: train_datagen = ImageDataGenerator(rescale= 1./255)
[17]: test_datagen = ImageDataGenerator(rescale= 1./255)
[18]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
      batch_size = 16
      train_datagen = ImageDataGenerator(
          rescale=1./255,
          horizontal_flip=True
      test_datagen = ImageDataGenerator(rescale=1./255)
      train_generator = train_datagen.flow_from_directory(
          train_path,
          target_size=(128, 128), # Smaller image size
          batch_size=batch_size, # Smaller batch size
                                  # Keep RGB if needed
          color_mode="rgb",
          class_mode="categorical"
      )
      test_generator = test_datagen.flow_from_directory(
          test_path,
          target_size=(128, 128), # Smaller image size
          batch_size=batch_size, # Smaller batch size
          color_mode="rgb",
          class_mode="categorical"
```

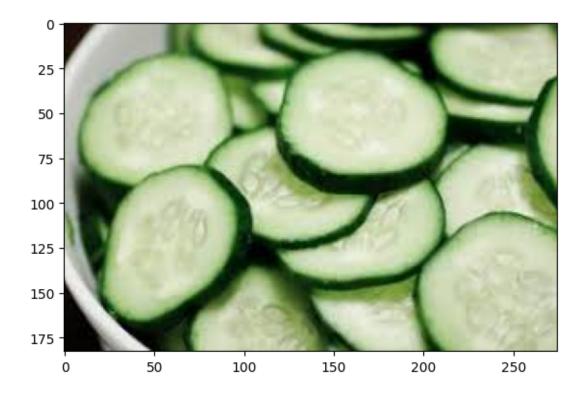
```
steps_per_epoch = train_generator.samples // batch_size
      validation_steps = test_generator.samples // batch_size
     Found 22564 images belonging to 2 classes.
     Found 2513 images belonging to 2 classes.
[19]: hist = model.fit(
          train_generator,
          epochs=10,
          validation_data=test_generator
     Epoch 1/10
     1411/1411
                           459s 323ms/step
     - accuracy: 0.7769 - loss: 0.4978 - val_accuracy: 0.8667 - val_loss: 0.3221
     Epoch 2/10
     1411/1411
                           478s 339ms/step
     - accuracy: 0.8419 - loss: 0.3899 - val_accuracy: 0.8544 - val_loss: 0.3297
     Epoch 3/10
     1411/1411
                           487s 345ms/step
     - accuracy: 0.8559 - loss: 0.3634 - val_accuracy: 0.9041 - val_loss: 0.2677
     Epoch 4/10
     1411/1411
                           526s 373ms/step
     - accuracy: 0.8713 - loss: 0.3365 - val_accuracy: 0.8794 - val_loss: 0.3210
     Epoch 5/10
     1411/1411
                           471s 334ms/step
     - accuracy: 0.8727 - loss: 0.3230 - val_accuracy: 0.8842 - val_loss: 0.3046
     Epoch 6/10
     1411/1411
                           437s 310ms/step
     - accuracy: 0.8805 - loss: 0.3125 - val_accuracy: 0.9160 - val_loss: 0.2324
     Epoch 7/10
     1411/1411
                           439s 311ms/step
     - accuracy: 0.8851 - loss: 0.2942 - val_accuracy: 0.9093 - val_loss: 0.2420
     Epoch 8/10
     1411/1411
                           445s 315ms/step
     - accuracy: 0.8922 - loss: 0.2724 - val_accuracy: 0.8910 - val_loss: 0.2925
     Epoch 9/10
     1411/1411
                           445s 315ms/step
     - accuracy: 0.8940 - loss: 0.2749 - val_accuracy: 0.9125 - val_loss: 0.2570
     Epoch 10/10
     1411/1411
                           437s 310ms/step
     - accuracy: 0.9070 - loss: 0.2480 - val_accuracy: 0.9077 - val_loss: 0.2577
[20]: plt.figure(figsize=[10,6])
      plt.plot(hist.history["accuracy"], label = "Train acc")
      plt.plot(hist.history["val_accuracy"], label = "Validation acc")
```

```
plt.legend()
plt.show()
```



```
[21]: plt.figure(figsize=(10,6))
    plt.plot(hist.history['loss'], label = "Train loss")
    plt.plot(hist.history['val_loss'], label = "Validation loss")
    plt.legend()
    plt.show()
```





[31]: test\_img = cv2.imread(r"C:\Users\Divish\Desktop\ml

→project\DATASET\TEST\R\R\_11017.jpg")

predict\_func(test\_img)



[]:[