Transfer Learning VGG 16 and VGG 19 using Keras

Please download the dataset from the below url

```
In [15]:
```

```
from tensorflow.compat.v1 import ConfigProto
from tensorflow.compat.v1 import InteractiveSession

config = ConfigProto()
config.gpu_options.per_process_gpu_memory_fraction = 0.5
config.gpu_options.allow_growth = True
session = InteractiveSession(config=config)
```

In [16]:

```
# import the libraries as shown below

from tensorflow.keras.layers import Input, Lambda, Dense, Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.applications.resnet50 import ResNet50
#from keras.applications.vgg16 import VGG16
from tensorflow.keras.applications.resnet50 import preprocess_input
from tensorflow.keras.preprocessing import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator,load_img
from tensorflow.keras.models import Sequential
import numpy as np
from glob import glob
#import matplotlib.pyplot as plt
```

In [17]:

```
# re-size all the images to this
IMAGE_SIZE = [224, 224]

train_path = 'Datasets/train'
valid_path = 'Datasets/test'
```

In [18]:

```
# Import the Vgg 16 library as shown below and add preprocessing layer to the front of VG
G
# Here we will be using imagenet weights
resnet = ResNet50(input_shape=IMAGE_SIZE + [3], weights='imagenet', include_top=False)
```

In [19]:

```
# don't train existing weights
for layer in resnet.layers:
    layer.trainable = False
```

In [20]:

```
# useful for getting number of output classes
folders = glob('Datasets/train/*')
```

In [21]:

```
# our layers - you can add more if you want
x = Flatten()(resnet.output)
```

In [22]:

```
prediction = Dense(len(folders), activation='softmax')(x)
```

```
# create a model object
model = Model(inputs=resnet.input, outputs=prediction)
```

In [23]:

view the structure of the model
model.summary()

Model: "model_1"

Model_1				
Layer (type)	Output	Shape	Param #	Connected to
input_2 (InputLayer)	[(None,	224, 224, 3)	0	
conv1_pad (ZeroPadding2D)	(None,	230, 230, 3)	0	input_2[0][0]
conv1_conv (Conv2D)	(None,	112, 112, 64)	9472	conv1_pad[0][0]
conv1_bn (BatchNormalization)	(None,	112, 112, 64)	256	conv1_conv[0][0]
conv1_relu (Activation)	(None,	112, 112, 64)	0	conv1_bn[0][0]
pool1_pad (ZeroPadding2D)	(None,	114, 114, 64)	0	conv1_relu[0][0]
pool1_pool (MaxPooling2D)	(None,	56, 56, 64)	0	pool1_pad[0][0]
conv2_block1_1_conv (Conv2D)	(None,	56, 56, 64)	4160	pool1_pool[0][0]
conv2_block1_1_bn (BatchNormali	(None,	56, 56, 64)	256	conv2_block1_1_conv[0][0
conv2_block1_1_relu (Activation	(None,	56, 56, 64)	0	conv2_block1_1_bn[0][0]
conv2_block1_2_conv (Conv2D)	(None,	56, 56, 64)	36928	conv2_block1_1_relu[0][0
conv2_block1_2_bn (BatchNormali]	(None,	56, 56, 64)	256	conv2_block1_2_conv[0][0
conv2_block1_2_relu (Activation	(None,	56, 56, 64)	0	conv2_block1_2_bn[0][0]
conv2_block1_0_conv (Conv2D)	(None,	56, 56, 256)	16640	pool1_pool[0][0]

<pre>conv2_block1_3_conv (Conv2D)]</pre>	(None,	56,	56,	256)	16640	conv2_block1_2_relu[0][0
conv2_block1_0_bn (BatchNormali	(None,	56,	56,	256)	1024	conv2_block1_0_conv[0][0
conv2_block1_3_bn (BatchNormali	(None,	56,	56,	256)	1024	conv2_block1_3_conv[0][0
conv2_block1_add (Add)	(None,	56,	56,	256)	0	conv2_block1_0_bn[0][0] conv2_block1_3_bn[0][0
	(27			05.6)		
conv2_block1_out (Activation)	(None,	56,	56,	256)	0	conv2_block1_add[0][0]
conv2_block2_1_conv (Conv2D)	(None,	56,	56,	64)	16448	conv2_block1_out[0][0]
conv2_block2_1_bn (BatchNormali]	(None,	56,	56,	64)	256	conv2_block2_1_conv[0][0
conv2_block2_1_relu (Activation	(None,	56,	56,	64)	0	conv2_block2_1_bn[0][0]
conv2_block2_2_conv (Conv2D)	(None,	56,	56,	64)	36928	conv2_block2_1_relu[0][0
conv2_block2_2_bn (BatchNormali	(None,	56,	56,	64)	256	conv2_block2_2_conv[0][0
conv2_block2_2_relu (Activation	(None,	56,	56,	64)	0	conv2_block2_2_bn[0][0]
conv2_block2_3_conv (Conv2D)	(None,	56,	56,	256)	16640	conv2_block2_2_relu[0][0
conv2_block2_3_bn (BatchNormali]	(None,	56,	56,	256)	1024	conv2_block2_3_conv[0][0
conv2_block2_add (Add)	(None,	56 ,	56,	256)	0	conv2_block1_out[0][0]
1						conv2_block2_3_bn[0][0
conv2_block2_out (Activation)	(None,	56,	56,	256)	0	conv2_block2_add[0][0]
conv2_block3_1_conv (Conv2D)	(None,	56,	56,	64)	16448	conv2_block2_out[0][0]
conv2_block3_1_bn (BatchNormali]	(None,	56,	56,	64)	256	conv2_block3_1_conv[0][0

conv2_block3_1_relu (Activation	(None,	56,	56,	64)	0	conv2_block3_1_bn[0][0]
conv2_block3_2_conv (Conv2D)	(None,	56,	56,	64)	36928	conv2_block3_1_relu[0][0
conv2_block3_2_bn (BatchNormali]	(None,	56,	56,	64)	256	conv2_block3_2_conv[0][0
conv2_block3_2_relu (Activation	(None,	56,	56,	64)	0	conv2_block3_2_bn[0][0]
conv2_block3_3_conv (Conv2D)	(None,	56,	56,	256)	16640	conv2_block3_2_relu[0][0
conv2_block3_3_bn (BatchNormali]	(None,	56,	56,	256)	1024	conv2_block3_3_conv[0][0
conv2_block3_add (Add)	(None,	56,	56,	256)	0	conv2_block2_out[0][0] conv2_block3_3_bn[0][0
conv2_block3_out (Activation)	(None,	56,	56,	256)	0	conv2_block3_add[0][0]
conv3_block1_1_conv (Conv2D)	(None,	28,	28,	128)	32896	conv2_block3_out[0][0]
conv3_block1_1_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block1_1_conv[0][0
conv3_block1_1_relu (Activation	(None,	28,	28,	128)	0	conv3_block1_1_bn[0][0]
conv3_block1_2_conv (Conv2D)	(None,	28,	28,	128)	147584	conv3_block1_1_relu[0][0
conv3_block1_2_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block1_2_conv[0][0
conv3_block1_2_relu (Activation	(None,	28,	28,	128)	0	conv3_block1_2_bn[0][0]
conv3_block1_0_conv (Conv2D)	(None,	28,	28,	512)	131584	conv2_block3_out[0][0]
conv3_block1_3_conv (Conv2D)	(None,	28,	28,	512)	66048	conv3_block1_2_relu[0][0
conv3_block1_0_bn (BatchNormali]	(None,	28,	28,	512)	2048	conv3_block1_0_conv[0][0
conv3_block1_3_bn (BatchNormali]	(None,	28,	28,	512)	2048	conv3_block1_3_conv[0][0

conv3_block1_add (Add)	(None,	28,	28,	512)	0	conv3_block1_0_bn[0][0]
1						conv3_block1_3_bn[0][0
conv3_block1_out (Activation)	(None,	28,	28,	512)	0	conv3_block1_add[0][0]
conv3_block2_1_conv (Conv2D)	(None,	28,	28,	128)	65664	conv3_block1_out[0][0]
conv3_block2_1_bn (BatchNormali	(None,	28,	28,	128)	512	conv3_block2_1_conv[0][0
conv3_block2_1_relu (Activation	(None,	28,	28,	128)	0	conv3_block2_1_bn[0][0]
conv3_block2_2_conv (Conv2D)	(None,	28,	28,	128)	147584	conv3_block2_1_relu[0][0
conv3_block2_2_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block2_2_conv[0][0
conv3_block2_2_relu (Activation	(None,	28,	28,	128)	0	conv3_block2_2_bn[0][0]
conv3_block2_3_conv (Conv2D)	(None,	28,	28,	512)	66048	conv3_block2_2_relu[0][0
<pre>conv3_block2_3_bn (BatchNormali]</pre>	(None,	28,	28,	512)	2048	conv3_block2_3_conv[0][0
conv3_block2_add (Add)	(None,	28,	28,	512)	0	conv3_block1_out[0][0]
1						conv3_block2_3_bn[0][0
conv3_block2_out (Activation)	(None,	28,	28,	512)	0	conv3_block2_add[0][0]
conv3_block3_1_conv (Conv2D)	(None,	28,	28,	128)	65664	conv3_block2_out[0][0]
conv3_block3_1_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block3_1_conv[0][0
conv3_block3_1_relu (Activation	(None,	28,	28,	128)	0	conv3_block3_1_bn[0][0]
conv3_block3_2_conv (Conv2D)	(None,	28,	28,	128)	147584	conv3_block3_1_relu[0][0
conv3_block3_2_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block3_2_conv[0][0

conv3_block3_2_relu (Activation	(None,	28,	28,	128)	0	conv3_block3_2_bn[0][0]
conv3_block3_3_conv (Conv2D)	(None,	28,	28,	512)	66048	conv3_block3_2_relu[0][0
conv3_block3_3_bn (BatchNormali]	(None,	28,	28,	512)	2048	conv3_block3_3_conv[0][0
conv3_block3_add (Add)	(None,	28,	28,	512)	0	conv3_block2_out[0][0]
]						conv3_block3_3_bn[0][0
conv3_block3_out (Activation)	(None,	28,	28,	512)	0	conv3_block3_add[0][0]
conv3_block4_1_conv (Conv2D)	(None,	28,	28,	128)	65664	conv3_block3_out[0][0]
conv3_block4_1_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block4_1_conv[0][0
conv3_block4_1_relu (Activation	(None,	28,	28,	128)	0	conv3_block4_1_bn[0][0]
conv3_block4_2_conv (Conv2D)	(None,	28,	28,	128)	147584	conv3_block4_1_relu[0][0
conv3_block4_2_bn (BatchNormali]	(None,	28,	28,	128)	512	conv3_block4_2_conv[0][0
conv3_block4_2_relu (Activation	(None,	28,	28,	128)	0	conv3_block4_2_bn[0][0]
conv3_block4_3_conv (Conv2D)	(None,	28,	28,	512)	66048	conv3_block4_2_relu[0][0
<pre>conv3_block4_3_bn (BatchNormali]</pre>	(None,	28,	28,	512)	2048	conv3_block4_3_conv[0][0
conv3_block4_add (Add)	(None,	28,	28,	512)	0	conv3_block3_out[0][0]
]						conv3_block4_3_bn[0][0
conv3_block4_out (Activation)	(None,	28,	28,	512)	0	conv3_block4_add[0][0]
conv4_block1_1_conv (Conv2D)	(None,	14,	14,	256)	131328	conv3_block4_out[0][0]
conv4_block1_1_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block1_1_conv[0][0

conv4_block1_1_relu (Activation	(None,	14,	14,	256)	0	conv4_block1_1_bn[0][0]
conv4_block1_2_conv (Conv2D)	(None,	14,	14,	256)	590080	conv4_block1_1_relu[0][0
conv4_block1_2_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block1_2_conv[0][0
conv4_block1_2_relu (Activation	(None,	14,	14,	256)	0	conv4_block1_2_bn[0][0]
conv4_block1_0_conv (Conv2D)	(None,	14,	14,	1024)	525312	conv3_block4_out[0][0]
conv4_block1_3_conv (Conv2D)	(None,	14,	14,	1024)	263168	conv4_block1_2_relu[0][0
<pre>conv4_block1_0_bn (BatchNormali]</pre>	(None,	14,	14,	1024)	4096	conv4_block1_0_conv[0][0
<pre>conv4_block1_3_bn (BatchNormali]</pre>	(None,	14,	14,	1024)	4096	conv4_block1_3_conv[0][0
conv4_block1_add (Add)	(None,	14,	14,	1024)	0	conv4_block1_0_bn[0][0]
]						conv4_block1_3_bn[0][0
conv4_block1_out (Activation)	(None,	14,	14,	1024)	0	conv4_block1_add[0][0]
conv4_block2_1_conv (Conv2D)	(None,	14,	14,	256)	262400	conv4_block1_out[0][0]
conv4_block2_1_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block2_1_conv[0][0
conv4_block2_1_relu (Activation	(None,	14,	14,	256)	0	conv4_block2_1_bn[0][0]
conv4_block2_2_conv (Conv2D)	(None,	14,	14,	256)	590080	conv4_block2_1_relu[0][0
conv4_block2_2_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block2_2_conv[0][0
conv4_block2_2_relu (Activation	(None,	14,	14,	256)	0	conv4_block2_2_bn[0][0]
conv4_block2_3_conv (Conv2D)	(None,	14,	14,	1024)	263168	conv4_block2_2_relu[0][0

	<pre>conv4_block2_3_bn (BatchNormali]</pre>	(None,	14,	14,	1024)	4096	conv4_block2_3_conv[0][0
Conv4_block2_out (Activation)	conv4_block2_add (Add)	(None,	14,	14,	1024)	0	
Conv4_block3_1_conv (Conv2D)]						conv4_block2_3_bn[0][0
Conv4_block3_1_bn (BatchNormali (None, 14, 14, 256) 1024 Conv4_block3_1_conv[0][0]	conv4_block2_out (Activation)	(None,	14,	14,	1024)	0	conv4_block2_add[0][0]
Conv4_block3_1_relu (Activation (None, 14, 14, 256) 0	conv4_block3_1_conv (Conv2D)	(None,	14,	14,	256)	262400	conv4_block2_out[0][0]
conv4_block3_2_conv (Conv2D) (None, 14, 14, 256) 590080 conv4_block3_1_relu[0][0] conv4_block3_2_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block3_2_conv[0][0] conv4_block3_2_relu (Activation (None, 14, 14, 256) 0 conv4_block3_2_bn[0][0] conv4_block3_3_conv (Conv2D) (None, 14, 14, 1024) 263168 conv4_block3_2_relu[0][0] conv4_block3_3_bn (BatchNormali (None, 14, 14, 1024) 4096 conv4_block3_3_conv[0][0] conv4_block3_add (Add) (None, 14, 14, 1024) 0 conv4_block3_3_bn[0][0] conv4_block3_out (Activation) (None, 14, 14, 1024) 0 conv4_block3_add[0][0] conv4_block4_1_conv (Conv2D) (None, 14, 14, 256) 262400 conv4_block3_out[0][0] conv4_block4_1_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block4_1_conv[0][0] conv4_block4_1_relu (Activation (None, 14, 14, 256) 0 conv4_block4_1_bn[0][0]	conv4_block3_1_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block3_1_conv[0][0
Conv4_block3_2_bn (BatchNormali (None, 14, 14, 256) 1024	conv4_block3_1_relu (Activation	(None,	14,	14,	256)	0	conv4_block3_1_bn[0][0]
Conv4_block3_2_relu (Activation (None, 14, 14, 256) 0	conv4_block3_2_conv (Conv2D)	(None,	14,	14,	256)	590080	conv4_block3_1_relu[0][0
conv4_block3_3_conv (Conv2D) (None, 14, 14, 1024) 263168 conv4_block3_2_relu[0][0 conv4_block3_3_bn (BatchNormali (None, 14, 14, 1024) 4096 conv4_block3_3_conv[0][0 conv4_block3_add (Add) (None, 14, 14, 1024) 0 conv4_block2_out[0][0] conv4_block3_add (Add) (None, 14, 14, 1024) 0 conv4_block3_abn[0][0 conv4_block3_out (Activation) (None, 14, 14, 1024) 0 conv4_block3_add[0][0] conv4_block4_1_conv (Conv2D) (None, 14, 14, 256) 262400 conv4_block3_out[0][0] conv4_block4_1_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block4_1_conv[0][0] conv4_block4_1_relu (Activation (None, 14, 14, 256) 0 conv4_block4_1_bn[0][0]	conv4_block3_2_bn (BatchNormali]	(None,	14,	14,	256)	1024	conv4_block3_2_conv[0][0
Conv4_block3_3_bn (BatchNormali (None, 14, 14, 1024) 4096	conv4_block3_2_relu (Activation	(None,	14,	14,	256)	0	conv4_block3_2_bn[0][0]
Conv4_block3_add (Add)	conv4_block3_3_conv (Conv2D)	(None,	14,	14,	1024)	263168	conv4_block3_2_relu[0][0
conv4_block3_3_bn[0][0 conv4_block3_out (Activation) (None, 14, 14, 1024) 0 conv4_block3_add[0][0] conv4_block4_1_conv (Conv2D) (None, 14, 14, 256) 262400 conv4_block3_out[0][0] conv4_block4_1_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block4_1_conv[0][0] conv4_block4_1_relu (Activation (None, 14, 14, 256) 0 conv4_block4_1_bn[0][0]	conv4_block3_3_bn (BatchNormali]	(None,	14,	14,	1024)	4096	conv4_block3_3_conv[0][0
	conv4_block3_add (Add)	(None,	14,	14,	1024)	0	conv4_block2_out[0][0]
conv4_block4_1_conv (Conv2D) (None, 14, 14, 256) 262400 conv4_block3_out[0][0] conv4_block4_1_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block4_1_conv[0][0] conv4_block4_1_relu (Activation (None, 14, 14, 256) 0 conv4_block4_1_bn[0][0]]						conv4_block3_3_bn[0][0
<pre>conv4_block4_1_bn (BatchNormali (None, 14, 14, 256) 1024 conv4_block4_1_conv[0][0] conv4_block4_1_relu (Activation (None, 14, 14, 256) 0 conv4_block4_1_bn[0][0]</pre>	conv4_block3_out (Activation)	(None,	14,	14,	1024)	0	conv4_block3_add[0][0]
	conv4_block4_1_conv (Conv2D)	(None,	14,	14,	256)	262400	conv4_block3_out[0][0]
	<pre>conv4_block4_1_bn (BatchNormali]</pre>	(None,	14,	14,	256)	1024	conv4_block4_1_conv[0][0
	conv4_block4_1_relu (Activation	(None,	14,	14,	256)	0	conv4_block4_1_bn[0][0]
	conv4_block4_2_conv (Conv2D)	(None,	14,	14,	256)	590080	conv4_block4_1_relu[0][0

(None,	14,	14,	256)	1024	conv4_block4_2_conv[0][0
n (None,	14,	14,	256)	0	conv4_block4_2_bn[0][0]
(None,	14,	14,	1024)	263168	conv4_block4_2_relu[0][0
(None,	14,	14,	1024)	4096	conv4_block4_3_conv[0][0
(None,	14,	14,	1024)	0	conv4_block3_out[0][0]
					conv4_block4_3_bn[0][0
(None,	14,	14,	1024)	0	conv4_block4_add[0][0]
(None,	14,	14,	256)	262400	conv4_block4_out[0][0]
(None,	14,	14,	256)	1024	conv4_block5_1_conv[0][0
n (None,	14,	14,	256)	0	conv4_block5_1_bn[0][0]
(None,	14,	14,	256)	590080	conv4_block5_1_relu[0][0
(None,	14,	14,	256)	1024	conv4_block5_2_conv[0][0
n (None,	14,	14,	256)	0	conv4_block5_2_bn[0][0]
(None,	14,	14,	1024)	263168	conv4_block5_2_relu[0][0
None,	14,	14,	1024)	4096	conv4_block5_3_conv[0][0
(None,	14,	14,	1024)	0	conv4_block4_out[0][0]
					conv4_block5_3_bn[0][0
(None,	14,	14,	1024)	0	conv4_block5_add[0][0]
(None,	14,	14,	256)	262400	conv4_block5_out[0][0]
	(None,	(None, 14, (None, 14,	(None, 14, 14, (None, 14, 14,	(None, 14, 14, 256) (None, 14, 14, 1024) (None, 14, 14, 1024) (None, 14, 14, 1024) (None, 14, 14, 256) (None, 14, 14, 1024) (None, 14, 14, 1024) (None, 14, 14, 1024)	

<pre>conv4_block6_1_bn (BatchNormali]</pre>	(None,	14, 14, 256)	1024	conv4_block6_1_conv[0][0
conv4_block6_1_relu (Activation	(None,	14, 14, 256)	0	conv4_block6_1_bn[0][0]
conv4_block6_2_conv (Conv2D)	(None,	14, 14, 256)	590080	conv4_block6_1_relu[0][0
conv4_block6_2_bn (BatchNormali	(None,	14, 14, 256)	1024	conv4_block6_2_conv[0][0
conv4_block6_2_relu (Activation	(None,	14, 14, 256)	0	conv4_block6_2_bn[0][0]
conv4_block6_3_conv (Conv2D)	(None,	14, 14, 1024)	263168	conv4_block6_2_relu[0][0
conv4_block6_3_bn (BatchNormali	(None,	14, 14, 1024)	4096	conv4_block6_3_conv[0][0
conv4_block6_add (Add)	(None,	14, 14, 1024)	0	conv4_block5_out[0][0]
1				conv4_block6_3_bn[0][0
conv4_block6_out (Activation)	(None,	14, 14, 1024)	0	conv4_block6_add[0][0]
conv5_block1_1_conv (Conv2D)	(None,	7, 7, 512)	524800	conv4_block6_out[0][0]
conv5_block1_1_bn (BatchNormali	(None,	7, 7, 512)	2048	conv5_block1_1_conv[0][0
conv5_block1_1_relu (Activation	(None,	7, 7, 512)	0	conv5_block1_1_bn[0][0]
conv5_block1_2_conv (Conv2D)	(None,	7, 7, 512)	2359808	conv5_block1_1_relu[0][0
conv5_block1_2_bn (BatchNormali	(None,	7, 7, 512)	2048	conv5_block1_2_conv[0][0
conv5_block1_2_relu (Activation	(None,	7, 7, 512)	0	conv5_block1_2_bn[0][0]
conv5_block1_0_conv (Conv2D)	(None,	7, 7, 2048)	2099200	conv4_block6_out[0][0]
conv5_block1_3_conv (Conv2D)	(None,	7, 7, 2048)	1050624	conv5_block1_2_relu[0][0
conv5_block1_0_bn (BatchNormali]	(None,	7, 7, 2048)	8192	conv5_block1_0_conv[0][0

<pre>conv5_block1_3_bn (BatchNormali]</pre>	(None,	7,	7,	2048)	8192	conv5_block1_3_conv[0][0
conv5_block1_add (Add)	(None,	7,	7,	2048)	0	conv5_block1_0_bn[0][0]
1						conv5_block1_3_bn[0][0
conv5_block1_out (Activation)	(None,	7,	7,	2048)	0	conv5_block1_add[0][0]
conv5_block2_1_conv (Conv2D)	(None,	7,	7,	512)	1049088	conv5_block1_out[0][0]
conv5_block2_1_bn (BatchNormali]	(None,	7,	7,	512)	2048	conv5_block2_1_conv[0][0
conv5_block2_1_relu (Activation	(None,	7,	7,	512)	0	conv5_block2_1_bn[0][0]
conv5_block2_2_conv (Conv2D)	(None,	7,	7,	512)	2359808	conv5_block2_1_relu[0][0
conv5_block2_2_bn (BatchNormali]	(None,	7,	7,	512)	2048	conv5_block2_2_conv[0][0
conv5_block2_2_relu (Activation	(None,	7,	7,	512)	0	conv5_block2_2_bn[0][0]
conv5_block2_3_conv (Conv2D)	(None,	7,	7,	2048)	1050624	conv5_block2_2_relu[0][0
conv5_block2_3_bn (BatchNormali]	(None,	7,	7,	2048)	8192	conv5_block2_3_conv[0][0
conv5_block2_add (Add)	(None,	7,	7,	2048)	0	conv5_block1_out[0][0]
1						conv5_block2_3_bn[0][0
conv5_block2_out (Activation)	(None,	7,	7,	2048)	0	conv5_block2_add[0][0]
conv5_block3_1_conv (Conv2D)	(None,	7,	7,	512)	1049088	conv5_block2_out[0][0]
conv5_block3_1_bn (BatchNormali	(None,	7,	7,	512)	2048	conv5_block3_1_conv[0][0
conv5_block3_1_relu (Activation	(None,	7,	7,	512)	0	conv5_block3_1_bn[0][0]
conv5_block3_2_conv (Conv2D)	(None,	7,	7,	512)	2359808	conv5_block3_1_relu[0][0

```
conv5 block3 2 bn (BatchNormali (None, 7, 7, 512)
                                                      2048
                                                                  conv5 block3 2 conv[0][0
1
conv5 block3 2 relu (Activation (None, 7, 7, 512)
                                                      0
                                                                   conv5 block3 2 bn[0][0]
conv5 block3_3_conv (Conv2D)
                              (None, 7, 7, 2048)
                                                      1050624
                                                                   conv5 block3 2 relu[0][0
conv5_block3_3_bn (BatchNormali (None, 7, 7, 2048)
                                                      8192
                                                                   conv5 block3 3 conv[0][0
conv5 block3 add (Add)
                                 (None, 7, 7, 2048)
                                                                   conv5 block2 out[0][0]
                                                                   conv5 block3 3 bn[0][0
]
conv5 block3 out (Activation)
                                 (None, 7, 7, 2048)
                                                      0
                                                                   conv5 block3 add[0][0]
flatten 1 (Flatten)
                                 (None, 100352)
                                                      0
                                                                   conv5 block3 out[0][0]
dense 1 (Dense)
                                                      401412
                                                                  flatten 1[0][0]
                                 (None, 4)
Total params: 23,989,124
Trainable params: 401,412
Non-trainable params: 23,587,712
```

In [24]:

```
# tell the model what cost and optimization method to use
model.compile(
  loss='categorical_crossentropy',
  optimizer='adam',
  metrics=['accuracy']
)
```

In [25]:

In [26]:

Found 1951 images helonging to 4 classes

In [27]:

Found 18 images belonging to 4 classes.

In [29]:

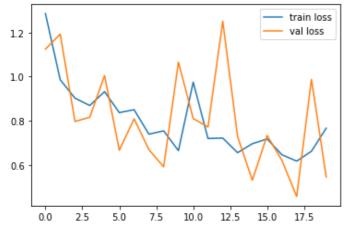
```
# fit the model
# Run the cell. It will take some time to execute
r = model.fit generator(
training set,
validation data=test set,
epochs=20,
steps per epoch=len(training set),
validation steps=len(test set)
Epoch 1/20
- val loss: 1.1249 - val accuracy: 0.5556
Epoch 2/20
- val loss: 1.1933 - val accuracy: 0.6111
Epoch 3/20
- val loss: 0.7966 - val accuracy: 0.7778
Epoch 4/20
- val loss: 0.8158 - val accuracy: 0.6667
Epoch 5/20
- val loss: 1.0051 - val accuracy: 0.6667
Epoch 6/20
loss: 0.6658 - val accuracy: 0.7778
- val
Epoch 7/20
- val loss: 0.8086 - val accuracy: 0.7222
Epoch 8/20
- val loss: 0.6687 - val accuracy: 0.7778
Epoch 9/20
- val_loss: 0.5906 - val_accuracy: 0.7778
Epoch 10/20
- val loss: 1.0657 - val accuracy: 0.7222
Epoch 11/20
- val loss: 0.8098 - val accuracy: 0.7778
Epoch 12/20
loss: 0.7715 - val accuracy: 0.7778
- val
Epoch 13/20
- val loss: 1.2519 - val accuracy: 0.7222
Epoch 14/20
- val loss: 0.7284 - val accuracy: 0.7778
Epoch 15/20
- val loss: 0.5299 - val accuracy: 0.8333
Epoch 16/20
- val_loss: 0.7328 - val_accuracy: 0.7778
Epoch 17/20
```

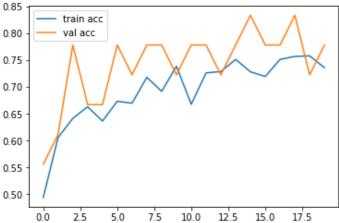
```
∪±/ ∪± [
                              TODD: 0.0100
                     100 27 11110/00CP
- val
   loss: 0.6202 - val accuracy: 0.7778
Epoch 18/20
- val_loss: 0.4560 - val_accuracy: 0.8333
Epoch 19/20
- val loss: 0.9876 - val accuracy: 0.7222
Epoch 20/20
- val loss: 0.5443 - val accuracy: 0.7778
In [32]:
import matplotlib.pyplot as plt
```

In [33]:

```
# plot the loss
plt.plot(r.history['loss'], label='train loss')
plt.plot(r.history['val_loss'], label='val loss')
plt.legend()
plt.show()
plt.savefig('LossVal_loss')

# plot the accuracy
plt.plot(r.history['accuracy'], label='train acc')
plt.plot(r.history['val_accuracy'], label='val acc')
plt.legend()
plt.show()
plt.savefig('AccVal_acc')
```





<Figure size 432x288 with 0 Axes>

In [34]:

```
# save it as a h5 file
from tensorflow.keras.models import load_model
```

```
model.save('model resnet50.h5')
In [ ]:
In [35]:
y pred = model.predict(test set)
In [36]:
y_pred
Out[36]:
array([[1.08071638e-03, 3.28080803e-02, 1.40938358e-02, 9.52017426e-01],
       [6.48212081e-05, 3.76272453e-07, 9.98108983e-01, 1.82576303e-03],
       [9.98069108e-01, 4.38354881e-08, 2.79694970e-04, 1.65126985e-03],
       [1.90574420e-03, 6.20834953e-07, 9.98076916e-01, 1.66221180e-05],
       [2.93544168e-03, 2.18342058e-03, 1.03987521e-02, 9.84482408e-01],
       [1.23585598e-03, 1.37195078e-04, 9.93926585e-01, 4.70046466e-03],
       [5.18485345e-03, 9.52562571e-01, 1.32831134e-04, 4.21197526e-02],
       [4.34744754e-04, 1.28285792e-02, 3.98747995e-03, 9.82749104e-01],
       [8.25350955e-02, 3.11684459e-01, 3.73537302e-01, 2.32243121e-01],
       [7.13894069e-01, 1.14700936e-01, 1.23989824e-02, 1.59005970e-01],
       [1.23955928e-01, 3.05530787e-01, 3.02462336e-02, 5.40267050e-01],
       [2.72632996e-03, 2.09897235e-02, 2.10788921e-02, 9.55205083e-01],
       [5.82802715e-03, 4.64445323e-01, 3.76655348e-02, 4.92061198e-01],
       [2.54945271e-03, 2.06118330e-06, 3.14365199e-04, 9.97134089e-01],
       [1.11564873e-02, 2.96992337e-04, 9.86988664e-01, 1.55796099e-03],
       [4.65637725e-03, 8.08736682e-01, 4.15630117e-02, 1.45043999e-01],
       [3.95378843e-03, 7.04863906e-01, 1.88800885e-04, 2.90993541e-01],
       [1.24013796e-02, 8.02336668e-04, 9.60396826e-01, 2.63994653e-02]],
      dtype=float32)
In [37]:
import numpy as np
y pred = np.argmax(y pred, axis=1)
In [38]:
y pred
Out[38]:
array([3, 2, 0, 2, 3, 2, 1, 3, 2, 0, 3, 3, 3, 3, 2, 1, 1, 2], dtype=int64)
In [ ]:
In [1]:
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
In [2]:
model=load model('model resnet50.h5')
In [39]:
img data
Out[39]:
array([[[ 6.7060997e+01, 5.4221001e+01, 4.7320000e+01],
         [ 6.9060997e+01, 5.6221001e+01, 4.9320000e+01],
         [7.3060997e+01, 6.0221001e+01, 5.3320000e+01],
```

```
[ 7.4060997e+01, 5.6221001e+01, 4.6320000e+01],
         [ 5.5060997e+01, 3.7221001e+01, 2.7320000e+01],
         [ 4.1060997e+01, 2.3221001e+01, 1.3320000e+01]],
        [[ 7.5060997e+01, 6.2221001e+01, 5.5320000e+01],
         [ 7.8060997e+01, 6.5221001e+01, 5.8320000e+01],
         [ 8.1060997e+01, 6.8221001e+01, 6.1320000e+01],
         [ 9.7060997e+01, 7.9221001e+01, 6.9320000e+01],
        [ 7.3060997e+01, 5.5221001e+01, 4.5320000e+01],
         [ 4.9060997e+01, 3.1221001e+01, 2.1320000e+01]],
        [[ 8.7060997e+01, 7.4221001e+01, 6.7320000e+01],
        [ 9.0060997e+01, 7.7221001e+01, 7.0320000e+01],
        [ 9.3060997e+01, 8.0221001e+01, 7.3320000e+01],
        [ 1.0106100e+02, 8.3221001e+01, 7.3320000e+01],
         [ 7.5060997e+01, 5.7221001e+01, 4.7320000e+01],
         [ 5.0060997e+01, 3.2221001e+01, 2.2320000e+01]],
        [[ 1.0406100e+02, 8.9221001e+01, 9.4320000e+01],
        [ 1.0206100e+02, 8.7221001e+01, 9.2320000e+01],
        [ 9.9060997e+01, 8.4221001e+01, 8.9320000e+01],
         [-1.0939003e+01, -1.6778999e+01, -1.4680000e+01],
         [-1.0939003e+01, -1.6778999e+01, -1.4680000e+01],
         [-1.0939003e+01, -1.6778999e+01, -1.4680000e+01]]
        [[ 1.0606100e+02, 9.1221001e+01, 9.6320000e+01],
        [ 1.0406100e+02, 8.9221001e+01, 9.4320000e+01],
        [ 1.0006100e+02, 8.5221001e+01, 9.0320000e+01],
        [-5.9390030e+00, -1.1778999e+01, -9.6800003e+00],
         [-5.9390030e+00, -1.1778999e+01, -9.6800003e+00],
         [-5.9390030e+00, -1.1778999e+01, -9.6800003e+00]]
        [[ 1.0806100e+02, 9.4221001e+01, 9.6320000e+01],
        [ 1.0606100e+02, 9.2221001e+01, 9.4320000e+01],
         [ 1.0206100e+02, 8.8221001e+01, 9.0320000e+01],
         [ 6.0997009e-02, -5.7789993e+00, -3.6800003e+00],
         [ 6.0997009e-02, -5.7789993e+00, -3.6800003e+00],
         [ 6.0997009e-02, -5.7789993e+00, -3.6800003e+00]]]],
     dtype=float32)
In [11]:
img=image.load img('Datasets/Test/Coffee/download (2).jpg',target size=(224,224))
In [12]:
x=image.img to array(img)
Out[12]:
array([[[254., 254., 254.],
       [254., 254., 254.],
        [254., 254., 254.],
        [254., 254., 254.],
        [255., 255., 255.],
       [255., 255., 255.]],
       [[254., 254., 254.],
        [254., 254., 254.],
        [254., 254., 254.],
        [254., 254., 254.],
```

```
[255., 255., 255.],
        [255., 255., 255.]],
       [[254., 254., 254.],
        [254., 254., 254.],
        [254., 254., 254.],
        [254., 254., 254.],
        [255., 255., 255.],
        [255., 255., 255.]],
       . . . ,
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        . . . ,
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        . . . ,
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]]], dtype=float32)
In [13]:
x.shape
Out[13]:
(224, 224, 3)
In [14]:
x=x/255
In [15]:
import numpy as np
x=np.expand dims(x,axis=0)
img_data=preprocess_input(x)
img data.shape
Out[15]:
(1, 224, 224, 3)
In [16]:
model.predict(img_data)
Out[16]:
array([[0.9745471, 0.0254529]], dtype=float32)
In [17]:
a=np.argmax(model.predict(img_data), axis=1)
```

```
In [102]:
    a==1
Out[102]:
    array([ True])
In [18]:
    import tensorflow as tf

In [19]:
    tf.__version__
Out[19]:
    '2.2.0'
In []:
```