Perform standard imports

```
! pip install split-folders
       Requirement already satisfied: split-folders in /usr/local/lib/python3.10/dist-packages (0.5.1)
In [2]: import tensorflow as tf
        from keras.models import Sequential
        from keras.layers import Activation, Dropout, Flatten, Dense, Conv2D, MaxPooling2D, BatchNormalization, GlobalAveragePooling2D
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        import splitfolders
        import numpy as np
        import copy
        import pandas as pd
        import matplotlib.pyplot as plt
        from matplotlib import style
        style.use('dark background')
In [3]: print('GPU name: ', tf.config.experimental.list physical devices('GPU'))
       GPU name: []
In [5]: gpus = tf.config.experimental.list physical devices('GPU')
        if gpus:
            try:
                for gpu in gpus:
                    tf.config.experimental.set memory growth(gpu, True)
                    logical gpus = tf.config.experimental.list logical devices('GPU')
                    print(len(gpus), "Physical GPUs,", len(logical gpus), "Logical GPUs")
            except RuntimeError as e:
                print(e)
In [ ]: from google.colab import drive
        drive.mount('/content/drive')
       Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force remount=Tru
       e).
```

```
In [ ]: %cd/content/drive/MyDrive/Course Material/Supervised Machine Learning/CNN_TF
    /content/drive/MyDrive/CNN_TF
```

Prepare train and test sets, loaders

```
In [6]: splitfolders.ratio("dataset", output="circle square",
            seed=1337, ratio=(.8, .2), group prefix=None, move=False) # default values
       Copying files: 200 files [00:00, 539.09 files/s]
In [7]: train datagen = ImageDataGenerator(rotation range=90,
                                              brightness range=[0.1, 0.7],
                                              horizontal flip=True,
                                              vertical flip=True,
                                              validation split=0.2,
                                              rescale = 1./255)
        test datagen = ImageDataGenerator(rescale = 1./255)
In [8]: batch size = 32
        training set = train datagen.flow from directory('circle square/train/',
                                                          target size = (200, 200),
                                                          class mode='binary',
                                                          shuffle = True,
                                                          batch size = batch size)
        test_set = train_datagen.flow_from_directory('circle_square/val/',
                                                          target size = (200, 200),
                                                          class mode='binary',
                                                          shuffle = False,
                                                          batch size = 1)
```

Found 160 images belonging to 2 classes. Found 40 images belonging to 2 classes.

Import ResNet-50

```
In [9]: base_model = tf.keras.applications.resnet50.ResNet50(weights='imagenet', include_top = False)
# setting include_top to False means the classification layers are not added and it is only a feature extracter and we will ad
# the clasification layers by ourself.
output = base_model.output

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50_weights_tf_dim_ordering_tf_k
ernels_notop.h5
94765736/94765736 6s @us/step
In [10]: base_model.summary()
```

Model: "resnet50"

Layer (type)	Output Shape	Param #	Connected to
input_layer (InputLayer)	(None, None, None, 3)	0	-
conv1_pad (ZeroPadding2D)	(None, None, None, 3)	0	input_layer[0][0]
conv1_conv (Conv2D)	(None, None, None, 64)	9,472	 conv1_pad[0][0]
conv1_bn (BatchNormalizatio	(None, None, None, 64)	256	 conv1_conv[0][0]
conv1_relu (Activation)	(None, None, None, 64)	0	conv1_bn[0][0]
pool1_pad (ZeroPadding2D)	(None, None, None, 64)	0	conv1_relu[0][0]
pool1_pool (MaxPooling2D)	(None, None, None, 64)	0	 pool1_pad[0][0]
conv2_block1_1_conv (Conv2D)	(None, None, None, 64)	4,160	 pool1_pool[0][0]
conv2_block1_1_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block1_1_c
conv2_block1_1_relu (Activation)	(None, None, None, 64)	0	conv2_block1_1_b
conv2_block1_2_conv (Conv2D)	(None, None, None, 64)	36,928	conv2_block1_1_r
conv2_block1_2_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block1_2_c
conv2_block1_2_relu (Activation)	(None, None, None, 64)	0	conv2_block1_2_b

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conv2_block1_0_conv (Conv2D)	(None, None, None, 256)	16,640	pool1_pool[0][0]
conv2_block1_3_conv (Conv2D)	(None, None, None, 256)	16,640	conv2_block1_2_r
conv2_block1_0_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv2_block1_0_c
conv2_block1_3_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv2_block1_3_c
conv2_block1_add (Add)	(None, None, None, 256)	0	conv2_block1_0_b conv2_block1_3_b
conv2_block1_out (Activation)	(None, None, None, 256)	0	conv2_block1_add
conv2_block2_1_conv (Conv2D)	(None, None, None, 64)	16,448	conv2_block1_out
conv2_block2_1_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block2_1_c
conv2_block2_1_relu (Activation)	(None, None, None, 64)	0	conv2_block2_1_b
conv2_block2_2_conv (Conv2D)	(None, None, None, 64)	36,928	conv2_block2_1_r
conv2_block2_2_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block2_2_c
conv2_block2_2_relu (Activation)	(None, None, None, 64)	0	conv2_block2_2_b
conv2_block2_3_conv (Conv2D)	(None, None, None, 256)	16,640	conv2_block2_2_r
conv2_block2_3_bn	(None, None,	1,024	conv2_block2_3_c

(BatchNormalizatio	None, 256)		
conv2_block2_add (Add)	(None, None, None, 256)	0	conv2_block1_out conv2_block2_3_b
conv2_block2_out (Activation)	(None, None, None, 256)	0	conv2_block2_add
conv2_block3_1_conv (Conv2D)	(None, None, None, 64)	16,448	conv2_block2_out
conv2_block3_1_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block3_1_c
conv2_block3_1_relu (Activation)	(None, None, None, 64)	0	conv2_block3_1_b
conv2_block3_2_conv (Conv2D)	(None, None, None, 64)	36,928	conv2_block3_1_r
conv2_block3_2_bn (BatchNormalizatio	(None, None, None, 64)	256	conv2_block3_2_c
conv2_block3_2_relu (Activation)	(None, None, None, 64)	0	conv2_block3_2_b
conv2_block3_3_conv (Conv2D)	(None, None, None, 256)	16,640	conv2_block3_2_r
conv2_block3_3_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv2_block3_3_c
conv2_block3_add (Add)	(None, None, None, 256)	0	conv2_block2_out conv2_block3_3_b
conv2_block3_out (Activation)	(None, None, None, 256)	0	conv2_block3_add
conv3_block1_1_conv (Conv2D)	(None, None, None, 128)	32,896	conv2_block3_out

conv3_block1_1_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block1_1_c
conv3_block1_1_relu (Activation)	(None, None, None, 128)	0	conv3_block1_1_b
conv3_block1_2_conv (Conv2D)	(None, None, None, 128)	147,584	conv3_block1_1_r
conv3_block1_2_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block1_2_c
conv3_block1_2_relu (Activation)	(None, None, None, 128)	0	conv3_block1_2_b
conv3_block1_0_conv (Conv2D)	(None, None, None, 512)	131,584	conv2_block3_out
conv3_block1_3_conv (Conv2D)	(None, None, None, 512)	66,048	conv3_block1_2_r
conv3_block1_0_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv3_block1_0_c
conv3_block1_3_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv3_block1_3_c
conv3_block1_add (Add)	(None, None, None, 512)	0	conv3_block1_0_b conv3_block1_3_b
conv3_block1_out (Activation)	(None, None, None, 512)	0	conv3_block1_add
conv3_block2_1_conv (Conv2D)	(None, None, None, 128)	65,664	conv3_block1_out
conv3_block2_1_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block2_1_c
conv3_block2_1_relu (Activation)	(None, None, None, 128)	0	conv3_block2_1_b

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conv3_block2_2_conv (Conv2D)	(None, None, None, 128)	147,584	conv3_block2_1_r
conv3_block2_2_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block2_2_c
conv3_block2_2_relu (Activation)	(None, None, None, 128)	0	conv3_block2_2_b
conv3_block2_3_conv (Conv2D)	(None, None, None, 512)	66,048	conv3_block2_2_r
conv3_block2_3_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv3_block2_3_c
conv3_block2_add (Add)	(None, None, None, 512)	0	conv3_block1_out conv3_block2_3_b
conv3_block2_out (Activation)	(None, None, None, 512)	0	conv3_block2_add
conv3_block3_1_conv (Conv2D)	(None, None, None, 128)	65,664	conv3_block2_out
conv3_block3_1_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block3_1_c
conv3_block3_1_relu (Activation)	(None, None, None, 128)	0	conv3_block3_1_b
conv3_block3_2_conv (Conv2D)	(None, None, None, 128)	147,584	conv3_block3_1_r
conv3_block3_2_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block3_2_c
conv3_block3_2_relu (Activation)	(None, None, None, 128)	0	conv3_block3_2_b
conv3_block3_3_conv	(None, None,	66,048	conv3_block3_2_r

(Conv2D)	None, 512)		
conv3_block3_3_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv3_block3_3_c
conv3_block3_add (Add)	(None, None, None, 512)	0	conv3_block2_out conv3_block3_3_b
conv3_block3_out (Activation)	(None, None, None, 512)	0	conv3_block3_add
conv3_block4_1_conv (Conv2D)	(None, None, None, 128)	65,664	conv3_block3_out
conv3_block4_1_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block4_1_c
conv3_block4_1_relu (Activation)	(None, None, None, 128)	0	conv3_block4_1_b
conv3_block4_2_conv (Conv2D)	(None, None, None, 128)	147,584	conv3_block4_1_r
conv3_block4_2_bn (BatchNormalizatio	(None, None, None, 128)	512	conv3_block4_2_c
conv3_block4_2_relu (Activation)	(None, None, None, 128)	0	conv3_block4_2_b
conv3_block4_3_conv (Conv2D)	(None, None, None, 512)	66,048	conv3_block4_2_r
conv3_block4_3_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv3_block4_3_c
conv3_block4_add (Add)	(None, None, None, 512)	0	conv3_block3_out conv3_block4_3_b
conv3_block4_out (Activation)	(None, None, None, 512)	0	conv3_block4_add

conv4_block1_1_conv (Conv2D)	(None, None, None, 256)	131,328	conv3_block4_out
conv4_block1_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block1_1_c
conv4_block1_1_relu (Activation)	(None, None, None, 256)	0	conv4_block1_1_b
conv4_block1_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block1_1_r
conv4_block1_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block1_2_c
conv4_block1_2_relu (Activation)	(None, None, None, 256)	0	conv4_block1_2_b
conv4_block1_0_conv (Conv2D)	(None, None, None, 1024)	525,312	conv3_block4_out
conv4_block1_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block1_2_r
conv4_block1_0_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block1_0_c
conv4_block1_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block1_3_c
conv4_block1_add (Add)	(None, None, None, 1024)	0	conv4_block1_0_b conv4_block1_3_b
conv4_block1_out (Activation)	(None, None, None, 1024)	0	conv4_block1_add
conv4_block2_1_conv (Conv2D)	(None, None, None, 256)	262,400	conv4_block1_out
conv4_block2_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block2_1_c

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<pre>conv4_block2_1_relu (Activation)</pre>	(None, None, None, 256)	0	conv4_block2_1_b
conv4_block2_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block2_1_r
conv4_block2_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block2_2_c
conv4_block2_2_relu (Activation)	(None, None, None, 256)	0	conv4_block2_2_b
conv4_block2_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block2_2_r
conv4_block2_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block2_3_c
conv4_block2_add (Add)	(None, None, None, 1024)	0	conv4_block1_out conv4_block2_3_b
conv4_block2_out (Activation)	(None, None, None, 1024)	0	conv4_block2_add
conv4_block3_1_conv (Conv2D)	(None, None, None, 256)	262,400	conv4_block2_out
conv4_block3_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block3_1_c
conv4_block3_1_relu (Activation)	(None, None, None, 256)	0	conv4_block3_1_b
conv4_block3_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block3_1_r
conv4_block3_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block3_2_c
conv4_block3_2_relu	(None, None,	0	conv4_block3_2_b

(Activation)	None, 256)		
conv4_block3_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block3_2_r
conv4_block3_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block3_3_c
conv4_block3_add (Add)	(None, None, None, 1024)	0	conv4_block2_out conv4_block3_3_b
conv4_block3_out (Activation)	(None, None, None, 1024)	0	conv4_block3_add
conv4_block4_1_conv (Conv2D)	(None, None, None, 256)	262,400	conv4_block3_out
conv4_block4_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block4_1_c
conv4_block4_1_relu (Activation)	(None, None, None, 256)	0	conv4_block4_1_b
conv4_block4_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block4_1_r
conv4_block4_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block4_2_c
conv4_block4_2_relu (Activation)	(None, None, None, 256)	0	conv4_block4_2_b
conv4_block4_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block4_2_r
conv4_block4_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block4_3_c
conv4_block4_add (Add)	(None, None, None, 1024)	0	conv4_block3_out conv4_block4_3_b

conv4_block4_out (Activation)	(None, None, None, 1024)	0	conv4_block4_add
conv4_block5_1_conv (Conv2D)	(None, None, None, 256)	262,400	conv4_block4_out
conv4_block5_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block5_1_c
conv4_block5_1_relu (Activation)	(None, None, None, 256)	0	conv4_block5_1_b
conv4_block5_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block5_1_r
conv4_block5_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block5_2_c
conv4_block5_2_relu (Activation)	(None, None, None, 256)	0	conv4_block5_2_b
conv4_block5_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block5_2_r
conv4_block5_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block5_3_c
conv4_block5_add (Add)	(None, None, None, 1024)	0	conv4_block4_out conv4_block5_3_b
conv4_block5_out (Activation)	(None, None, None, 1024)	0	conv4_block5_add
conv4_block6_1_conv (Conv2D)	(None, None, None, 256)	262,400	conv4_block5_out
conv4_block6_1_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block6_1_c
conv4_block6_1_relu (Activation)	(None, None, None, 256)	0	conv4_block6_1_b

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conv4_block6_2_conv (Conv2D)	(None, None, None, 256)	590,080	conv4_block6_1_r
conv4_block6_2_bn (BatchNormalizatio	(None, None, None, 256)	1,024	conv4_block6_2_c
conv4_block6_2_relu (Activation)	(None, None, None, 256)	0	conv4_block6_2_b
conv4_block6_3_conv (Conv2D)	(None, None, None, 1024)	263,168	conv4_block6_2_r
conv4_block6_3_bn (BatchNormalizatio	(None, None, None, 1024)	4,096	conv4_block6_3_c
conv4_block6_add (Add)	(None, None, None, 1024)	0	conv4_block5_out conv4_block6_3_b
conv4_block6_out (Activation)	(None, None, None, 1024)	0	conv4_block6_add
conv5_block1_1_conv (Conv2D)	(None, None, None, 512)	524,800	conv4_block6_out
conv5_block1_1_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block1_1_c
conv5_block1_1_relu (Activation)	(None, None, None, 512)	0	conv5_block1_1_b
conv5_block1_2_conv (Conv2D)	(None, None, None, 512)	2,359,808	conv5_block1_1_r
conv5_block1_2_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block1_2_c
conv5_block1_2_relu (Activation)	(None, None, None, 512)	0	conv5_block1_2_b
conv5_block1_0_conv	(None, None,	2,099,200	conv4_block6_out

(Conv2D)	None, 2048)		
conv5_block1_3_conv (Conv2D)	(None, None, None, 2048)	1,050,624	conv5_block1_2_r
conv5_block1_0_bn (BatchNormalizatio	(None, None, None, 2048)	8,192	conv5_block1_0_c
conv5_block1_3_bn (BatchNormalizatio	(None, None, None, 2048)	8,192	conv5_block1_3_c
conv5_block1_add (Add)	(None, None, None, 2048)	0	conv5_block1_0_b conv5_block1_3_b
conv5_block1_out (Activation)	(None, None, None, 2048)	0	conv5_block1_add
conv5_block2_1_conv (Conv2D)	(None, None, None, 512)	1,049,088	conv5_block1_out
conv5_block2_1_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block2_1_c
conv5_block2_1_relu (Activation)	(None, None, None, 512)	0	conv5_block2_1_b
conv5_block2_2_conv (Conv2D)	(None, None, None, 512)	2,359,808	conv5_block2_1_r
conv5_block2_2_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block2_2_c
conv5_block2_2_relu (Activation)	(None, None, None, 512)	0	conv5_block2_2_b
conv5_block2_3_conv (Conv2D)	(None, None, None, 2048)	1,050,624	conv5_block2_2_r
conv5_block2_3_bn (BatchNormalizatio	(None, None, None, 2048)	8,192	conv5_block2_3_c

conv5_block2_add (Add)	(None, None, None, 2048)	0	conv5_block1_out conv5_block2_3_b
conv5_block2_out (Activation)	(None, None, None, 2048)	0	conv5_block2_add
conv5_block3_1_conv (Conv2D)	(None, None, None, 512)	1,049,088	conv5_block2_out
conv5_block3_1_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block3_1_c
conv5_block3_1_relu (Activation)	(None, None, None, 512)	0	conv5_block3_1_b
conv5_block3_2_conv (Conv2D)	(None, None, None, 512)	2,359,808	conv5_block3_1_r
conv5_block3_2_bn (BatchNormalizatio	(None, None, None, 512)	2,048	conv5_block3_2_c
conv5_block3_2_relu (Activation)	(None, None, None, 512)	0	conv5_block3_2_b
conv5_block3_3_conv (Conv2D)	(None, None, None, 2048)	1,050,624	conv5_block3_2_r
conv5_block3_3_bn (BatchNormalizatio	(None, None, None, 2048)	8,192	conv5_block3_3_c
conv5_block3_add (Add)	(None, None, None, 2048)	0	conv5_block2_out conv5_block3_3_b
conv5_block3_out (Activation)	(None, None, None, 2048)	0	conv5_block3_add

Total params: 23,587,712 (89.98 MB)

Trainable params: 23,534,592 (89.78 MB)
Non-trainable params: 53,120 (207.50 KB)

Freeze upto 143 layers

```
In [11]: for layer in base_model.layers[:143]: layer.trainable = False
In [12]: for i, layer in enumerate(base_model.layers):
    print(i, layer.name,"-", layer.trainable)
```

0 input layer - False 1 conv1 pad - False 2 conv1 conv - False 3 conv1 bn - False 4 conv1 relu - False 5 pool1 pad - False 6 pool1 pool - False 7 conv2 block1 1 conv - False 8 conv2 block1 1 bn - False 9 conv2 block1 1 relu - False 10 conv2 block1 2 conv - False 11 conv2 block1 2 bn - False 12 conv2 block1 2 relu - False 13 conv2 block1 0 conv - False 14 conv2 block1 3 conv - False 15 conv2 block1 0 bn - False 16 conv2 block1 3 bn - False 17 conv2 block1_add - False 18 conv2 block1 out - False 19 conv2 block2 1 conv - False 20 conv2 block2 1 bn - False 21 conv2 block2 1 relu - False 22 conv2 block2 2 conv - False 23 conv2 block2 2 bn - False 24 conv2 block2 2 relu - False 25 conv2 block2 3 conv - False 26 conv2_block2_3_bn - False 27 conv2 block2 add - False 28 conv2 block2 out - False 29 conv2 block3 1 conv - False 30 conv2 block3 1 bn - False 31 conv2_block3_1_relu - False 32 conv2 block3 2 conv - False 33 conv2_block3_2_bn - False 34 conv2 block3 2 relu - False 35 conv2 block3 3 conv - False 36 conv2 block3 3 bn - False 37 conv2 block3 add - False 38 conv2 block3 out - False 39 conv3 block1 1 conv - False 40 conv3 block1 1 bn - False

41 conv3 block1 1 relu - False 42 conv3 block1 2 conv - False 43 conv3 block1 2 bn - False 44 conv3 block1 2 relu - False 45 conv3 block1 0 conv - False 46 conv3 block1 3 conv - False 47 conv3 block1 0 bn - False 48 conv3 block1 3 bn - False 49 conv3 block1 add - False 50 conv3 block1 out - False 51 conv3 block2 1 conv - False 52 conv3 block2 1 bn - False 53 conv3 block2 1 relu - False 54 conv3 block2 2 conv - False 55 conv3 block2 2 bn - False 56 conv3 block2 2 relu - False 57 conv3 block2 3 conv - False 58 conv3 block2 3 bn - False 59 conv3 block2 add - False 60 conv3 block2 out - False 61 conv3 block3 1 conv - False 62 conv3 block3 1 bn - False 63 conv3 block3 1 relu - False 64 conv3 block3 2 conv - False 65 conv3 block3 2 bn - False 66 conv3 block3 2 relu - False 67 conv3 block3 3 conv - False 68 conv3 block3 3 bn - False 69 conv3 block3 add - False 70 conv3 block3 out - False 71 conv3 block4 1 conv - False 72 conv3 block4_1_bn - False 73 conv3 block4 1 relu - False 74 conv3_block4_2_conv - False 75 conv3 block4 2 bn - False 76 conv3 block4 2 relu - False 77 conv3 block4 3 conv - False 78 conv3 block4 3 bn - False 79 conv3 block4 add - False 80 conv3 block4 out - False 81 conv4 block1 1 conv - False 82 conv4 block1 1 bn - False 83 conv4 block1 1 relu - False 84 conv4 block1 2 conv - False 85 conv4 block1_2_bn - False 86 conv4 block1 2 relu - False 87 conv4 block1 0 conv - False 88 conv4 block1 3 conv - False 89 conv4 block1 0 bn - False 90 conv4 block1 3 bn - False 91 conv4 block1 add - False 92 conv4 block1 out - False 93 conv4 block2 1 conv - False 94 conv4 block2 1 bn - False 95 conv4 block2 1 relu - False 96 conv4 block2 2 conv - False 97 conv4 block2 2 bn - False 98 conv4 block2 2 relu - False 99 conv4 block2 3 conv - False 100 conv4 block2 3 bn - False 101 conv4 block2 add - False 102 conv4 block2 out - False 103 conv4 block3 1 conv - False 104 conv4 block3 1 bn - False 105 conv4 block3 1 relu - False 106 conv4 block3 2 conv - False 107 conv4 block3 2 bn - False 108 conv4 block3 2 relu - False 109 conv4 block3 3 conv - False 110 conv4 block3 3 bn - False 111 conv4 block3 add - False 112 conv4 block3 out - False 113 conv4 block4_1_conv - False 114 conv4 block4 1 bn - False 115 conv4 block4 1 relu - False 116 conv4 block4 2 conv - False 117 conv4 block4 2 bn - False 118 conv4 block4 2 relu - False 119 conv4 block4 3 conv - False 120 conv4 block4 3 bn - False 121 conv4 block4 add - False 122 conv4 block4 out - False

123 conv4 block5 1 conv - False 124 conv4 block5 1 bn - False 125 conv4 block5 1 relu - False 126 conv4 block5 2 conv - False 127 conv4 block5 2 bn - False 128 conv4 block5 2 relu - False 129 conv4 block5 3 conv - False 130 conv4 block5 3 bn - False 131 conv4 block5 add - False 132 conv4 block5 out - False 133 conv4 block6 1 conv - False 134 conv4 block6 1 bn - False 135 conv4 block6 1 relu - False 136 conv4 block6 2 conv - False 137 conv4 block6 2 bn - False 138 conv4 block6 2 relu - False 139 conv4 block6 3 conv - False 140 conv4 block6 3 bn - False 141 conv4 block6 add - False 142 conv4 block6 out - False 143 conv5_block1_1_conv - True 144 conv5 block1 1 bn - True 145 conv5 block1 1 relu - True 146 conv5 block1 2 conv - True 147 conv5 block1 2 bn - True 148 conv5 block1 2 relu - True 149 conv5 block1 0 conv - True 150 conv5 block1 3 conv - True 151 conv5 block1 0 bn - True 152 conv5 block1 3 bn - True 153 conv5 block1 add - True 154 conv5 block1_out - True 155 conv5 block2 1 conv - True 156 conv5 block2 1 bn - True 157 conv5 block2 1 relu - True 158 conv5 block2 2 conv - True 159 conv5 block2 2 bn - True 160 conv5 block2 2 relu - True 161 conv5 block2 3 conv - True 162 conv5 block2 3 bn - True 163 conv5 block2 add - True

```
164 conv5_block2_out - True
165 conv5_block3_1_conv - True
166 conv5_block3_1_bn - True
167 conv5_block3_1_relu - True
168 conv5_block3_2_conv - True
169 conv5_block3_2_bn - True
170 conv5_block3_2_relu - True
171 conv5_block3_3_conv - True
172 conv5_block3_3_bn - True
173 conv5_block3_add - True
174 conv5_block3 out - True
```

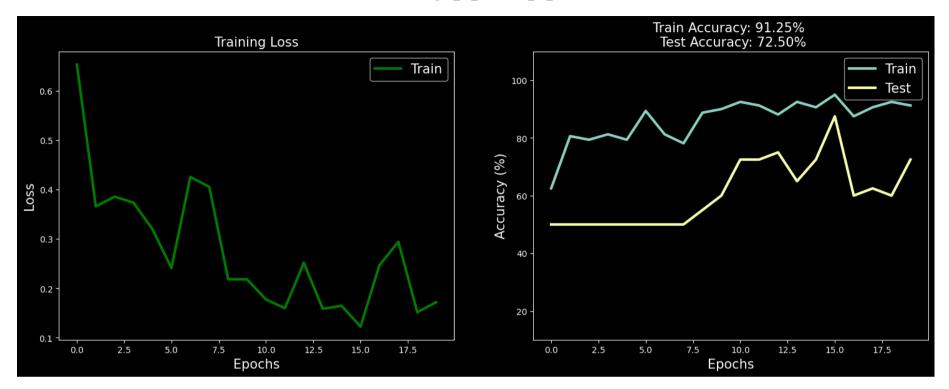
Adding more layers to base model

```
In [13]: model = Sequential()
    model.add(base_model)
    model.add(tf.keras.layers.GlobalAveragePooling2D()) # For Flattening
    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(1, activation='sigmoid'))
```

Training and Testing the model

Epoch		
5/5 — Epoch		Os 5s/step - accuracy: 0.5881 - loss: 0.6863 - val_accuracy: 0.5000 - val_loss: 0.6973
5/5 —		3s 5s/step - accuracy: 0.7839 - loss: 0.3975 - val_accuracy: 0.5000 - val_loss: 0.6992
Epoch 5/5 —		3s 5s/step - accuracy: 0.8193 - loss: 0.3336 - val accuracy: 0.5000 - val loss: 0.7023
Epoch		
5/5 — Epoch		3s 5s/step - accuracy: 0.8134 - loss: 0.3530 - val_accuracy: 0.5000 - val_loss: 0.6964
5/5 -	-	3s 5s/step - accuracy: 0.7542 - loss: 0.3505 - val_accuracy: 0.5000 - val_loss: 0.6893
Epoch		25 55 /54 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
5/5 — Epoch		3s 5s/step - accuracy: 0.8986 - loss: 0.2298 - val_accuracy: 0.5000 - val_loss: 0.6882
5/5 -		3s 5s/step - accuracy: 0.7795 - loss: 0.4719 - val_accuracy: 0.5000 - val_loss: 0.6893
Epoch 5/5 —		3s 5s/step - accuracy: 0.7600 - loss: 0.4445 - val_accuracy: 0.5000 - val_loss: 0.6878
Epoch	9/20	
5/5 -	10/20	3s 5s/step - accuracy: 0.8731 - loss: 0.2251 - val_accuracy: 0.5500 - val_loss: 0.6821
5/5 -	•	3s 5s/step - accuracy: 0.9298 - loss: 0.1943 - val_accuracy: 0.6000 - val_loss: 0.6795
•	11/20	3s 5s/step - accuracy: 0.9364 - loss: 0.1691 - val accuracy: 0.7250 - val loss: 0.6723
5/5 — Epoch	12/20	55 35/Step - accuracy. 0.9304 - 1055. 0.1091 - Val_accuracy. 0.7230 - Val_1055. 0.6723
5/5 -		2s 5s/step - accuracy: 0.9183 - loss: 0.1550 - val_accuracy: 0.7250 - val_loss: 0.6700
5/5 —	13/20	3s 5s/step - accuracy: 0.8884 - loss: 0.2270 - val_accuracy: 0.7500 - val_loss: 0.6661
	14/20	
5/5 — Epoch	15/20	3s 5s/step - accuracy: 0.9082 - loss: 0.1710 - val_accuracy: 0.6500 - val_loss: 0.6609
5/5 —		5s 5s/step - accuracy: 0.8845 - loss: 0.1957 - val_accuracy: 0.7250 - val_loss: 0.6622
Epoch 5/5 —	16/20	3s 5s/step - accuracy: 0.9425 - loss: 0.1326 - val accuracy: 0.8750 - val loss: 0.6433
	17/20	23 33/300p accuracy. 0.3423 1033. 0.1320 var_accuracy. 0.0730 var_1033. 0.0433
5/5 -		3s 5s/step - accuracy: 0.8885 - loss: 0.2290 - val_accuracy: 0.6000 - val_loss: 0.6442
5/5 -	18/20	2s 5s/step - accuracy: 0.9232 - loss: 0.2294 - val_accuracy: 0.6250 - val_loss: 0.6396
•	19/20	10 Fo/ston
5/5 — Epoch	20/20	!s 5s/step - accuracy: 0.9329 - loss: 0.1362 - val_accuracy: 0.6000 - val_loss: 0.6293
5/5 —		2s 5s/step - accuracy: 0.9027 - loss: 0.1810 - val_accuracy: 0.7250 - val_loss: 0.5989

```
In [16]: trainAcc = [100 * x for x in history.history['accuracy']]
         testAcc = [100 * x for x in history.history['val accuracy']]
In [17]: fig,ax = plt.subplots(1,2,figsize=(18,6))
         ax[0].plot(history.history['loss'], 'g', lw = 3, label = 'Train')
         ax[0].set xlabel('Epochs', fontsize = 15)
         ax[0].set ylabel('Loss', fontsize = 15)
         ax[0].legend(fontsize = 15)
         ax[0].set title('Training Loss', fontsize = 15)
         ax[1].plot(trainAcc, label ='Train', lw = 3)
         ax[1].plot(testAcc, label ='Test', lw = 3)
         ax[1].set xlabel('Epochs', fontsize = 15)
         ax[1].set ylabel('Accuracy (%)', fontsize = 15)
         ax[1].set ylim([10,110])
         ax[1].set title(f'Train Accuracy: {trainAcc[-1]:.2f}% \n Test Accuracy: {testAcc[-1]:.2f}%', fontsize = 15)
         ax[1].legend(fontsize = 15)
         plt.show()
```



```
import tensorflow as tf
from keras.models import Sequential
from keras.layers import Activation, Dropout, Flatten, Dense, Conv2D, MaxPooling2D
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import splitfolders
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import style
style.use('dark_background')
```

Prepare train and test sets, loaders

```
Copying files: 200 files [00:00, 533.33 files/s]
```

Creating train and test data generators

```
In [22]: train_datagen = ImageDataGenerator(rescale = 1./255)
    test_datagen = ImageDataGenerator(rescale = 1./255)

In [23]: num_classes = 2
    img_shape = (200,200,3)
    batch_size = 32
```

Creating Training and Testing Datasets

Found 160 images belonging to 2 classes.

Found 40 images belonging to 2 classes.

Create CNN

```
In [26]: model = Sequential()
  model.add(Conv2D(10 ,kernel_size = (5, 5), padding = 'SAME', input_shape = img_shape))
  model.add(Activation('relu'))
  model.add(MaxPooling2D(pool_size=(2, 2),strides = 2))
```

```
model.add(Conv2D(20, (5, 5), padding='SAME')) # "SAME" tries to pad evenly left and right
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2, 2),strides = 2))

model.add(Flatten())
model.add(Dense(64))
model.add(Activation('relu'))
model.add(Dense(1))
model.add(Activation('sigmoid'))

C:\Users\siddh\AppData\Roaming\Python\Python39\site-packages\keras\src\layers\convolutional\base_conv.py:107: UserWarning: Do n
ot pass an `input_shape'/input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object a
s the first layer in the model instead.
super().__init__(activity_regularizer=activity_regularizer, **kwargs)

In [27]: model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 200, 200, 10)	760
activation (Activation)	(None, 200, 200, 10)	0
max_pooling2d (MaxPooling2D)	(None, 100, 100, 10)	0
conv2d_1 (Conv2D)	(None, 100, 100, 20)	5,020
activation_1 (Activation)	(None, 100, 100, 20)	0
max_pooling2d_1 (MaxPooling2D)	(None, 50, 50, 20)	0
flatten (Flatten)	(None, 50000)	0
dense_2 (Dense)	(None, 64)	3,200,064
activation_2 (Activation)	(None, 64)	0
dense_3 (Dense)	(None, 1)	65
activation_3 (Activation)	(None, 1)	0

Total params: 3,205,909 (12.23 MB)

Trainable params: 3,205,909 (12.23 MB)

Non-trainable params: 0 (0.00 B)

Setting Early Stopping Criterion

Training and Testing the model

```
- 3s 386ms/step - accuracy: 0.4552 - loss: 1.1016 - val accuracy: 0.5000 - val loss: 0.8297
        5/5 ---
        Epoch 2/500
                                - 2s 312ms/step - accuracy: 0.5218 - loss: 0.7595 - val accuracy: 0.5000 - val loss: 0.6904
        5/5 -
        Epoch 3/500
        5/5 -
                                - 2s 328ms/step - accuracy: 0.5923 - loss: 0.6895 - val accuracy: 0.6750 - val loss: 0.6838
        Epoch 4/500
        5/5 ---
                                - 2s 363ms/step - accuracy: 0.5990 - loss: 0.6841 - val accuracy: 0.7000 - val loss: 0.6772
        Epoch 5/500
                                - 2s 324ms/step - accuracy: 0.7078 - loss: 0.6723 - val accuracy: 0.7000 - val loss: 0.6627
        5/5 ---
        Epoch 6/500
        5/5 -
                                 2s 337ms/step - accuracy: 0.6490 - loss: 0.6654 - val accuracy: 0.7000 - val loss: 0.6494
        Epoch 7/500
        5/5 -
                                - 2s 313ms/step - accuracy: 0.7002 - loss: 0.6468 - val accuracy: 0.7000 - val loss: 0.6528
        Epoch 8/500
                                 2s 329ms/step - accuracy: 0.7217 - loss: 0.6289 - val accuracy: 0.7250 - val loss: 0.6107
        5/5 -
        Epoch 9/500
        5/5 -
                                - 2s 356ms/step - accuracy: 0.7745 - loss: 0.5584 - val accuracy: 0.6750 - val loss: 0.5411
        Epoch 10/500
                                - 2s 345ms/step - accuracy: 0.7902 - loss: 0.4827 - val accuracy: 0.7000 - val loss: 0.6525
        5/5 ---
        Epoch 11/500
        5/5 -
                                - 2s 380ms/step - accuracy: 0.7530 - loss: 0.4981 - val accuracy: 0.7000 - val loss: 0.5611
        Epoch 12/500

    2s 390ms/step - accuracy: 0.8387 - loss: 0.4459 - val accuracy: 0.7250 - val loss: 0.5804

        Epoch 12: early stopping
        Restoring model weights from the end of the best epoch: 9.
In [31]: trainAcc = [100 * x for x in hist.history['accuracy']]
         testAcc = [100 * x for x in hist.history['val accuracy']]
```

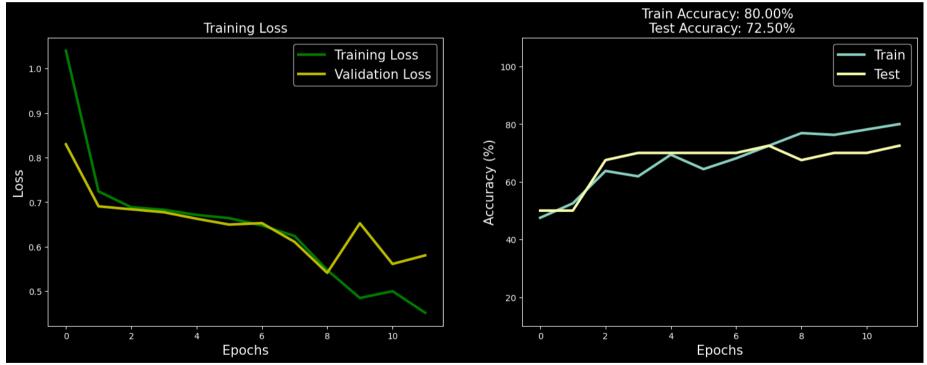
Visualize the Model Performance

```
In [32]: fig,ax = plt.subplots(1,2,figsize=(18,6))

ax[0].plot(hist.history['loss'], 'g', lw = 3, label = 'Training Loss')
ax[0].plot(hist.history['val_loss'], 'y', lw = 3, label = 'Validation Loss')
ax[0].set_xlabel('Epochs', fontsize = 15)
ax[0].set_ylabel('Loss', fontsize = 15)
ax[0].legend(fontsize = 15)
```

```
ax[0].set_title('Training Loss', fontsize = 15)

ax[1].plot(trainAcc, label ='Train', lw = 3)
ax[1].plot(testAcc, label ='Test', lw = 3)
ax[1].set_xlabel('Epochs', fontsize = 15)
ax[1].set_ylabel('Accuracy (%)', fontsize = 15)
ax[1].set_ylim([10,110])
ax[1].set_title(f'Train Accuracy: {trainAcc[-1]:.2f}% \n Test Accuracy: {testAcc[-1]:.2f}%', fontsize = 15)
ax[1].legend(fontsize = 15)
plt.show()
```



In []: