```
In [3]: from google.colab import drive
        drive.mount('/content/gdrive')
        Mounted at /content/gdrive
In [4]:
       !unzip gdrive/MyDrive/birds/test_data.zip
                  gdrive/MyDrive/birds/test data.zip
        Archive:
           creating: test data/
           creating: test_data/blasti/
          inflating: test data/blasti/DSC 6396.jpg
          inflating: test_data/blasti/DSC_6397.jpg
          inflating: test_data/blasti/DSC_6398.jpg
          inflating: test_data/blasti/DSC_6399.jpg
          inflating: test_data/blasti/DSC_6400.jpg
          inflating: test data/blasti/DSC 6401.jpg
          inflating: test_data/blasti/DSC_6402.jpg
          inflating: test_data/blasti/DSC_6403.jpg
          inflating: test_data/blasti/DSC_6405.jpg
          inflating: test_data/blasti/DSC_6406.jpg
          inflating: test data/blasti/DSC 6407.jpg
          inflating: test_data/blasti/DSC_6408.jpg
          inflating: test data/blasti/DSC 6409.jpg
          inflating: test_data/blasti/DSC_6410.jpg
          inflating: test_data/blasti/DSC_6411.jpg
           creating: test_data/boneql/
In [5]: |!unzip gdrive/MyDrive/birds/train_data.zip
                  gdrive/MyDrive/birds/train data.zip
        Archive:
           creating: train data/
           creating: train_data/blasti/
          inflating: train_data/blasti/DSC_6382.jpg
          inflating: train_data/blasti/DSC_6383.jpg
          inflating: train_data/blasti/DSC_6384-2.jpg
          inflating: train data/blasti/DSC 6384.jpg
          inflating: train_data/blasti/DSC_6385.jpg
          inflating: train data/blasti/DSC 6386.jpg
          inflating: train_data/blasti/DSC_6387.jpg
          inflating: train_data/blasti/DSC_6388.jpg
          inflating: train_data/blasti/DSC_6389.jpg
          inflating: train_data/blasti/DSC_6390.jpg
          inflating: train data/blasti/DSC 6391.jpg
          inflating: train_data/blasti/DSC_6392.jpg
          inflating: train_data/blasti/DSC_6393.jpg
          inflating: train_data/blasti/DSC_6394.jpg
          inflating: train_data/blasti/DSC_6395.jpg
           creating: train_data/bonegl/
```

```
In [6]: from tensorflow.keras.layers import Dense, Flatten, Input
         from tensorflow.keras.models import Model
         from tensorflow.keras.preprocessing import image
         from tensorflow.keras.preprocessing.image import ImageDataGenerator
         import numpy as np
 In [7]: train_path = '/content/train_data'
         test path = '/content/test data'
 In [8]: train_gen = ImageDataGenerator(rescale=1./255,
                                        shear range=0.2,
                                        zoom_range=0.2,
                                        horizontal flip=True)
         test_gen = ImageDataGenerator(rescale=1./255)
 In [9]: train = train_gen.flow_from_directory(train_path,
                                               target_size=(224,224),
                                               batch_size=22,
                                               class_mode='categorical')
         test = test_gen.flow_from_directory(test_path,
                                               target size=(224,224),
                                               batch size=22.
                                               class_mode='categorical')
         Found 150 images belonging to 16 classes.
         Found 157 images belonging to 16 classes.
         ##VGG16
In [10]: from tensorflow.keras.applications.vgg16 import VGG16, preprocess_i
In [11]: vqq = VGG16(include_top=False, weights='imagenet', input_shape=(224,2)
         Downloading data from https://storage.googleapis.com/tensorflow/ke
         ras-applications/vgg16/vgg16 weights tf dim ordering tf kernels no
         top.h5 (https://storage.googleapis.com/tensorflow/keras-applicatio
         ns/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5)
         58889256/58889256 [============== ] - 2s @us/step
```

```
In [12]: for layer in vgg.layers:
    print(layer)
```

<keras.engine.input\_layer.InputLayer object at 0x7f56f7406320> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f7407be0</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f7448b50</pre> <keras.layers.pooling.max pooling2d.MaxPooling2D object at 0x7f56f</pre> 7448c40> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f7449870</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f744a6e0</pre> <keras.layers.pooling.max pooling2d.MaxPooling2D object at 0x7f56f</pre> 744b760> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f744ac50</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f744bfd0</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f7448580</pre> <keras.layers.pooling.max pooling2d.MaxPooling2D object at 0x7f56f</pre> 5364f70> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f53661a0</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f5367010</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f5366d70</pre> <keras.layers.pooling.max\_pooling2d.MaxPooling2D object at 0x7f56f</pre> 5380640> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f5365f90</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f5366710</pre> <keras.layers.convolutional.conv2d.Conv2D object at 0x7f56f5382320</pre> <keras.layers.pooling.max pooling2d.MaxPooling2D object at 0x7f56f</pre> 53833d0>

```
In [13]: for layer in vgg.layers:
    layer.trainable=False
```

```
In [14]: x = Flatten()(vgg.output)
```

```
In [15]: prediction = Dense(16,activation='softmax')(x)
```

```
In [16]: model = Model(inputs=vgg.input,outputs=prediction)
```

In [17]:

## model.summary()

Model: "model"

block1_conv1 (Conv2D) (	[(None, 224, 224, 3)] (None, 224, 224, 64) (None, 224, 224, 64) (None, 112, 112, 64)	0 1792 36928
_	(None, 224, 224, 64)	
block1_conv2 (Conv2D) (		36928
	(None, 112, 112, 64)	
block1_pool (MaxPooling2D) (		0
block2_conv1 (Conv2D) (	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D) (	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D) (	(None, 56, 56, 128)	0
block3_conv1 (Conv2D) (	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D) (	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D) (	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D) (	(None, 28, 28, 256)	0
block4_conv1 (Conv2D) (	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D) (	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D) (	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D) (	(None, 14, 14, 512)	0
block5_conv1 (Conv2D) (	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D) (	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D) (	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D) (	(None, 7, 7, 512)	0
flatten (Flatten) (	(None, 25088)	0
dense (Dense) (	(None, 16)	401424

Total params: 15,116,112
Trainable params: 401,424

Non-trainable params: 14,714,688

```
In [18]: model.compile(loss='categorical_crossentropy',optimizer='adam',metr
In [19]: model.fit generator(train, validation data=test, epochs=10, steps per
        <ipython-input-19-c19f13cb6f36>:1: UserWarning: `Model.fit_generat
        or` is deprecated and will be removed in a future version. Please
        use `Model.fit`, which supports generators.
         model.fit_generator(train,validation_data=test,epochs=10,steps_p
        er_epoch=len(train), validation_steps=len(test))
        Epoch 1/10
        7/7 [=============== ] - 108s 15s/step - loss: 3.963
        1 - accuracy: 0.1400 - val_loss: 3.2677 - val_accuracy: 0.1847
        Epoch 2/10
        7/7 [============== ] - 91s 14s/step - loss: 1.9688
        - accuracy: 0.4600 - val_loss: 2.9406 - val_accuracy: 0.2484
        Epoch 3/10
        - accuracy: 0.6867 - val loss: 3.1222 - val accuracy: 0.3185
        Epoch 4/10
        9 - accuracy: 0.7933 - val loss: 2.8847 - val accuracy: 0.3631
        Epoch 5/10
        7/7 [============== ] - 92s 15s/step - loss: 0.3683
        - accuracy: 0.9000 - val_loss: 2.9352 - val_accuracy: 0.3503
        Epoch 6/10
        7/7 [============== ] - 88s 14s/step - loss: 0.2516
        - accuracy: 0.9467 - val_loss: 2.9953 - val_accuracy: 0.3567
        Epoch 7/10
        7/7 [============== ] - 92s 15s/step - loss: 0.1970
        - accuracy: 0.9467 - val loss: 3.0808 - val accuracy: 0.3694
        Epoch 8/10
        7/7 [============== ] - 93s 15s/step - loss: 0.1391
        - accuracy: 0.9733 - val loss: 3.1347 - val accuracy: 0.3694
        Epoch 9/10
        7/7 [=============== ] - 91s 14s/step - loss: 0.0923
        - accuracy: 1.0000 - val_loss: 2.8206 - val_accuracy: 0.4204
        Epoch 10/10
        8 - accuracy: 1.0000 - val_loss: 2.8776 - val_accuracy: 0.4140
Out[19]: <keras.callbacks.History at 0x7f56f40d2e60>
        ##ResNet50
In [20]: from tensorflow.keras.applications.resnet50 import ResNet50
```

```
In [21]: resnet = ResNet50(include_top=False,input_shape=(224,224,3))
         Downloading data from https://storage.googleapis.com/tensorflow/ke
         ras-applications/resnet/resnet50 weights tf dim ordering tf kernel
         s_notop.h5 (https://storage.googleapis.com/tensorflow/keras-applic
         ations/resnet/resnet50_weights_tf_dim_ordering_tf_kernels_notop.h5
         In [22]: for layer in resnet.layers:
           print(layer)
         <keras.engine.input_layer.InputLayer object at 0x7f56f40d3bb0>
         <keras.layers.reshaping.zero padding2d.ZeroPadding2D object at 0x7</pre>
         f56f539a950>
         <keras.layers.convolutional.conv2d.Conv2D object at 0x7f5667f17d60</pre>
         <keras.layers.normalization.batch normalization.BatchNormalization</pre>
         object at 0x7f56f539bb20>
         <keras.layers.core.activation.Activation object at 0x7f5667f17820>
         <keras.layers.reshaping.zero_padding2d.ZeroPadding2D object at 0x7</pre>
         f5667f17970>
         <keras.layers.pooling.max_pooling2d.MaxPooling2D object at 0x7f566</pre>
         <keras.layers.convolutional.conv2d.Conv2D object at 0x7f5667f26ec0</pre>
         <keras.layers.normalization.batch normalization.BatchNormalization</pre>
         object at 0x7f5667f27dc0>
         <keras.layers.core.activation.Activation object at 0x7f5667f24f70>
         <keras.layers.convolutional.conv2d.Conv2D object at 0x7f5667f252a0</pre>
                           ualiantian batak magmaliantian DatakNagmaliantian
In [23]: | x = Flatten()(resnet.output)
In [24]: | out = Dense(16, activation='softmax')(x)
In [25]:
         res model = Model(inputs=resnet.input,outputs=out)
```

```
In [26]: res_model.summary()
         Model: "model_1"
          Layer (type)
                                          Output Shape
                                                               Param #
                                                                            C
         onnected to
          input_2 (InputLayer)
                                          [(None, 224, 224, 3 0
                                          )]
          conv1_pad (ZeroPadding2D)
                                          (None, 230, 230, 3)
         'input_2[0][0]']
          conv1_conv (Conv2D)
                                          (None, 112, 112, 64 9472
         'conv1_pad[0][0]']
                                          )
          conv1_bn (BatchNormalization)
                                         (None, 112, 112, 64 256
In [27]: res_model.compile(loss='categorical_crossentropy',optimizer='adam'
```

```
Epoch 1/10
62 - accuracy: 0.2000 - val_loss: 101.7114 - val_accuracy: 0.0573
Epoch 2/10
7/7 [============== ] - 92s 14s/step - loss: 8.1270
- accuracy: 0.4933 - val_loss: 1438412.2500 - val_accuracy: 0.0573
Epoch 3/10
7/7 [=============== ] - 92s 14s/step - loss: 7.5458
- accuracy: 0.4533 - val_loss: 5777523.0000 - val_accuracy: 0.1274
Epoch 4/10
7/7 [============= ] - 122s 19s/step - loss: 4.022
0 - accuracy: 0.5467 - val loss: 7797352.5000 - val accuracy: 0.05
Epoch 5/10
7/7 [=============== ] - 91s 14s/step - loss: 6.7724
- accuracy: 0.6333 - val_loss: 2681508.5000 - val_accuracy: 0.1274
Epoch 6/10
7/7 [============= ] - 90s 14s/step - loss: 10.926
3 - accuracy: 0.5067 - val_loss: 10694477.0000 - val_accuracy: 0.1
274
Epoch 7/10
7/7 [=============== ] - 93s 14s/step - loss: 2.5144
- accuracy: 0.5667 - val_loss: 128072608.0000 - val_accuracy: 0.12
74
Epoch 8/10
- accuracy: 0.6333 - val_loss: 52614392.0000 - val_accuracy: 0.127
4
Epoch 9/10
7/7 [=============== ] - 90s 14s/step - loss: 1.8227
- accuracy: 0.6800 - val_loss: 355240640.0000 - val_accuracy: 0.12
74
Epoch 10/10
7/7 [============ ] - 93s 14s/step - loss: 3.5036
- accuracy: 0.6867 - val_loss: 752114048.0000 - val_accuracy: 0.12
```

Out[28]: <keras.callbacks.History at 0x7f5667e4e4a0>

## VGG16 gives higher val\_accuracy.

```
In [29]: import numpy as np
from tensorflow.keras.preprocessing import image
```