In [1]:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt
from PIL import Image
import os
```

In [2]:

git clone https://github.com/harshad12patre/crop_yield_prediction.git

```
Cloning into 'crop_yield_prediction'...
remote: Enumerating objects: 19804, done.
remote: Counting objects: 100% (19804/19804), done.
remote: Compressing objects: 100% (19774/19774), done.
remote: Total 19804 (delta 53), reused 19773 (delta 26), pack-reused 0
Receiving objects: 100% (19804/19804), 338.72 MiB | 42.78 MiB/s, done.
Resolving deltas: 100% (53/53), done.
Checking out files: 100% (19708/19708), done.
```

In [12]:

```
pretrained_model =
tf.keras.applications.VGG19(include_top=True, weights='imagenet', input_shape=(224,
224,3))
pretrained_model.summary()
```

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 224, 224, 3)]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	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_conv4 (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_conv4 (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_conv4 (Conv2D)	(None,	14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None,	7, 7, 512)	0
flatten (Flatten)	(None,	25088)	0
fc1 (Dense)	(None,	4096)	102764544
fc2 (Dense)	(None,	4096)	16781312
predictions (Dense)	(None,	1000)	4097000

Total params: 143,667,240
Trainable params: 143,667,240

Non-trainable params: 0

In [13]:

pretrained_model =
tf.keras.applications.VGG19(include_top=False,weights='imagenet',input_shape=(224,224,3))
pretrained_model.summary()

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 224, 224, 3)]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	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_conv4 (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_conv4 (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_conv4 (Conv2D)	(None,	14,	14,	512)	2359808
block5_pool (MaxPooling2D)	(None,	7,	7, 5	12)	0
Total params: 20,024,384 Trainable params: 20,024,384 Non-trainable params: 0				=	

Non-trainable params: 0

In [5]:

```
for layer in pretrained model.layers:
  layer.trainable = False
```

In [14]:

```
last layer = pretrained model.get layer("block5 pool")
output = last layer.output
X = tf.keras.layers.Flatten()(output)
X = tf.keras.layers.Dense(1024,activation="relu")(X)
X = tf.keras.layers.Dense(512,activation="relu")(X)
X = tf.keras.layers.Dense(13,activation="softmax")(X)
model = tf.keras.models.Model(inputs=pretrained_model.input,outputs=X)
tf.keras.backend.clear session()
model.compile(loss="categorical crossentropy",optimizer="adam",metrics=["accuracy"]
])
```

In [10]:

```
train datagen = ImageDataGenerator(rescale=1/255.0, rotation range=20,
      width shift range=0.1,
     height shift range=0.1,
     shear range=0.1,
     zoom range=0.1,
     horizontal_flip=True,
     fill mode='nearest',
     validation split=0.2)
batch size = 128
train_generator =
train_datagen.flow_from_directory("/content/crop_yield_prediction/color"
```

```
,batch_size=128,target_size=(2
,224), class mode="categorical", subset="training")
validation generator =
train datagen.flow from directory("/content/crop yield prediction/color"
                                    ,batch size=128,target size=(2
,224),class mode="categorical",subset="validation")
history = model.fit(train generator, steps per epoch=train generator.samples//batc
h size, validation data=validation generator, validation steps=validation generator
.samples//batch size,epochs=5,verbose=1)
Found 15691 images belonging to 13 classes.
Found 3920 images belonging to 13 classes.
Epoch 1/5
: 0.5964 - val loss: 0.1390 - val accuracy: 0.9604
Epoch 2/5
: 0.9680 - val loss: 0.1055 - val accuracy: 0.9677
Epoch 3/5
: 0.9747 - val loss: 0.0899 - val accuracy: 0.9745
: 0.9804 - val loss: 0.0646 - val accuracy: 0.9792
Epoch 5/5
: 0.9858 - val loss: 0.0527 - val accuracy: 0.9857
In [11]:
loss = history.history['loss']
```

```
accuracy = history.history['accuracy']
val loss = history.history['val loss']
val accuracy = history.history['val accuracy']
epoch = range(len(loss))
plt.figure()
plt.plot(epoch, loss)
plt.plot(epoch, val loss)
plt.xlabel("Epoch#")
plt.ylabel("Accuracy/Loss")
plt.title("Loss")
plt.show()
plt.figure()
plt.plot(epoch,accuracy)
plt.plot(epoch, val accuracy)
plt.xlabel("Epoch#")
plt.ylabel("Val Accuracy/Loss")
plt.title("Accuracy")
plt.show()
```



