

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 [3]:

```
pretrained_model =
tf.keras.applications.VGG16(include_top=True, weights='imagenet', input_shape=(224,
224, 3))
pretrained_model.summary()
```

```
Downloading data from https://storage.googleapis.com/tensorflow/keras-
applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467904/553467096 [=====] - 3s 0us/step
Model: "vgg16"
```

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_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
fc1 (Dense)	(None, 4096)	102764544
fc2 (Dense)	(None, 4096)	16781312
predictions (Dense)	(None, 1000)	4097000
=====		
Total params: 138,357,544		
Trainable params: 138,357,544		
Non-trainable params: 0		

In [4]:

```
pretrained_model =
tf.keras.applications.VGG16(include_top=False,weights='imagenet',input_shape=(224
,224,3))
pretrained_model.summary()
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5
58892288/58889256 [=====] - 0s 0us/step
Model: "vgg16"

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_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
=====		
Total params: 14,714,688		
Trainable params: 14,714,688		
Non-trainable params: 0		

In [5]:

```
for layer in pretrained_model.layers:
    layer.trainable = False
```

In [9]:

```
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//batch_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

```
Epoch 1/5
122/122 [=====] - 230s 2s/step - loss: 2.8797 - accuracy
: 0.5964 - val_loss: 0.1390 - val_accuracy: 0.9604
Epoch 2/5
122/122 [=====] - 226s 2s/step - loss: 0.1131 - accuracy
: 0.9680 - val_loss: 0.1055 - val_accuracy: 0.9677
Epoch 3/5
122/122 [=====] - 226s 2s/step - loss: 0.0783 - accuracy
: 0.9747 - val_loss: 0.0899 - val_accuracy: 0.9745
Epoch 4/5
122/122 [=====] - 227s 2s/step - loss: 0.0600 - accuracy
: 0.9804 - val_loss: 0.0646 - val_accuracy: 0.9792
Epoch 5/5
122/122 [=====] - 226s 2s/step - loss: 0.0473 - accuracy
: 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()
```



