

```
In [2]: import pandas as pd
import numpy as np
import keras
import os
import matplotlib.pyplot as plt
from keras.preprocessing import image
import random
import pickle
from keras import models, layers, callbacks
from keras.callbacks import ModelCheckpoint
from keras.models import load_model
import tensorflow as tf
from tensorflow.keras.layers import Input, Lambda, Dense, Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.applications.inception_v3 import InceptionV3
from tensorflow.keras.applications.inception_v3 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 shutil
import cv2
from math import sqrt, floor
from prettytable import PrettyTable
```

```

In [3]: classes= []
        sample_counts= []

        for f in os.listdir(r'D:\seed Classification Model\train'):
            train_class_path= os.path.join(r'D:\seed Classification Model\train', f)
            if os.path.isdir(train_class_path):
                classes.append(f)
                sample_counts.append(len(os.listdir(train_class_path)))

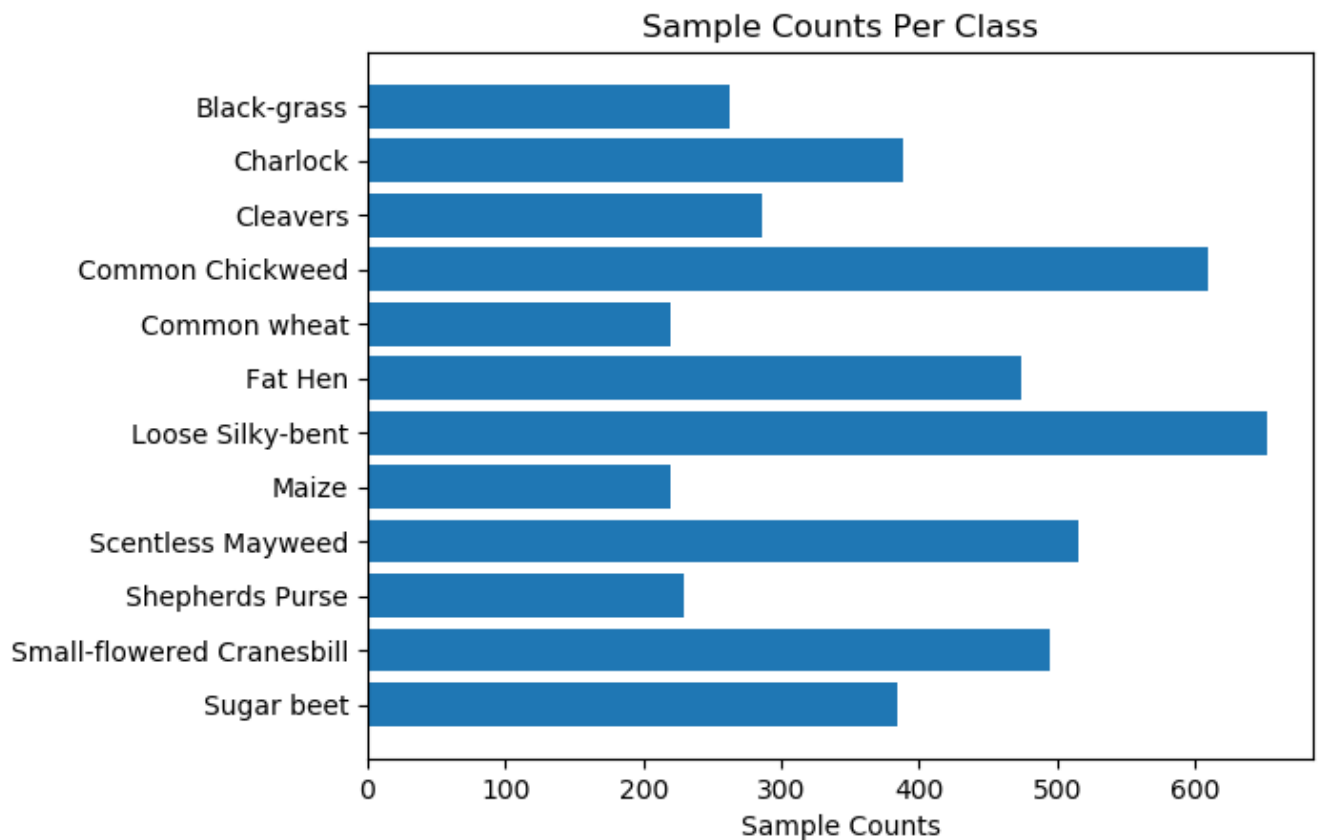
        plt.rcdefaults()
        fig, ax = plt.subplots()

        # Example data
        y_pos = np.arange(len(classes))

        ax.barh(y_pos, sample_counts, align='center')
        ax.set_yticks(y_pos)
        ax.set_yticklabels(classes)
        ax.invert_yaxis() # labels read top-to-bottom
        ax.set_xlabel('Sample Counts')
        ax.set_title('Sample Counts Per Class')

        plt.show()

```



```

In [4]: #create validation set
def create_validation(validation_split=0.2):
    if os.path.isdir('validation'):
        print('Validation directory already created!')
        print('Process Terminated')
        return

    os.mkdir(r'D:\seed Classification Model\validation')
    for f in os.listdir(r'D:\seed Classification Model\train'):
        train_class_path= os.path.join(r'D:\seed Classification Model\train',
f)
        if os.path.isdir(train_class_path):
            validation_class_path= os.path.join(r'D:\seed Classification Model
\validation', f)
            os.mkdir(validation_class_path)
            files_to_move= int(0.2*len(os.listdir(train_class_path)))
            random_image= os.path.join(train_class_path, random.choice(os.listdir(train_class_path)))
            shutil.move(random_image, validation_class_path)
        print('Validation set created successfully using {:.2%} of training data'.format(validation_split))

```

```

In [5]: create_validation()

```

```

Validation directory already created!
Process Terminated

```

```

In [6]: sample_counts= {}

for i, d in enumerate([r'D:\seed Classification Model\train', r'D:\seed Classification Model\validation']):

    classes= []
    sample_counts[d]= []

    for f in os.listdir(d):
        train_class_path= os.path.join(d, f)
        if os.path.isdir(train_class_path):
            classes.append(f)
            sample_counts[d].append(len(os.listdir(train_class_path)))

    #fig, ax= plt.subplot(221+i)
    fig, ax = plt.subplots()

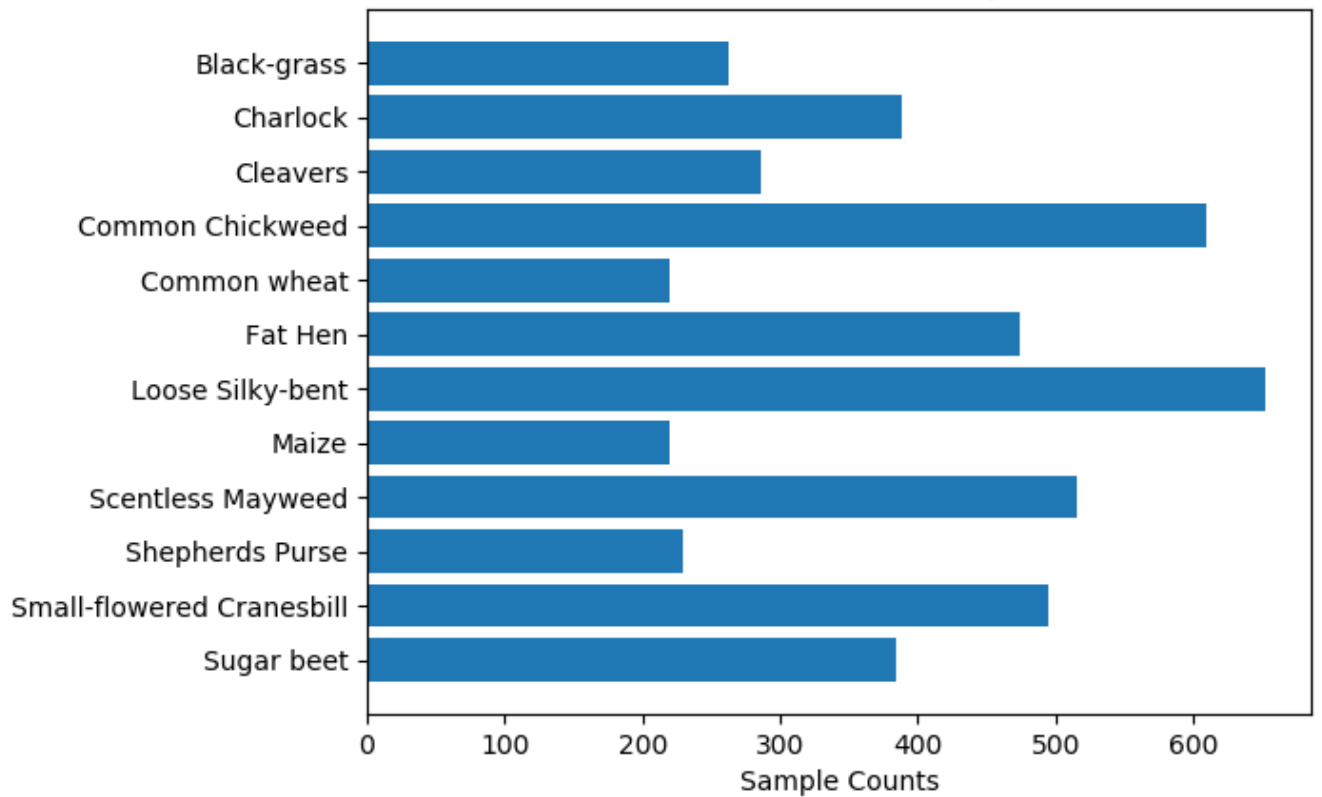
    # Example data
    y_pos = np.arange(len(classes))

    ax.barh(y_pos, sample_counts[d], align='center')
    ax.set_yticks(y_pos)
    ax.set_yticklabels(classes)
    ax.invert_yaxis() # labels read top-to-bottom
    ax.set_xlabel('Sample Counts')
    ax.set_title('{} Sample Counts Per Class'.format(d.capitalize()))

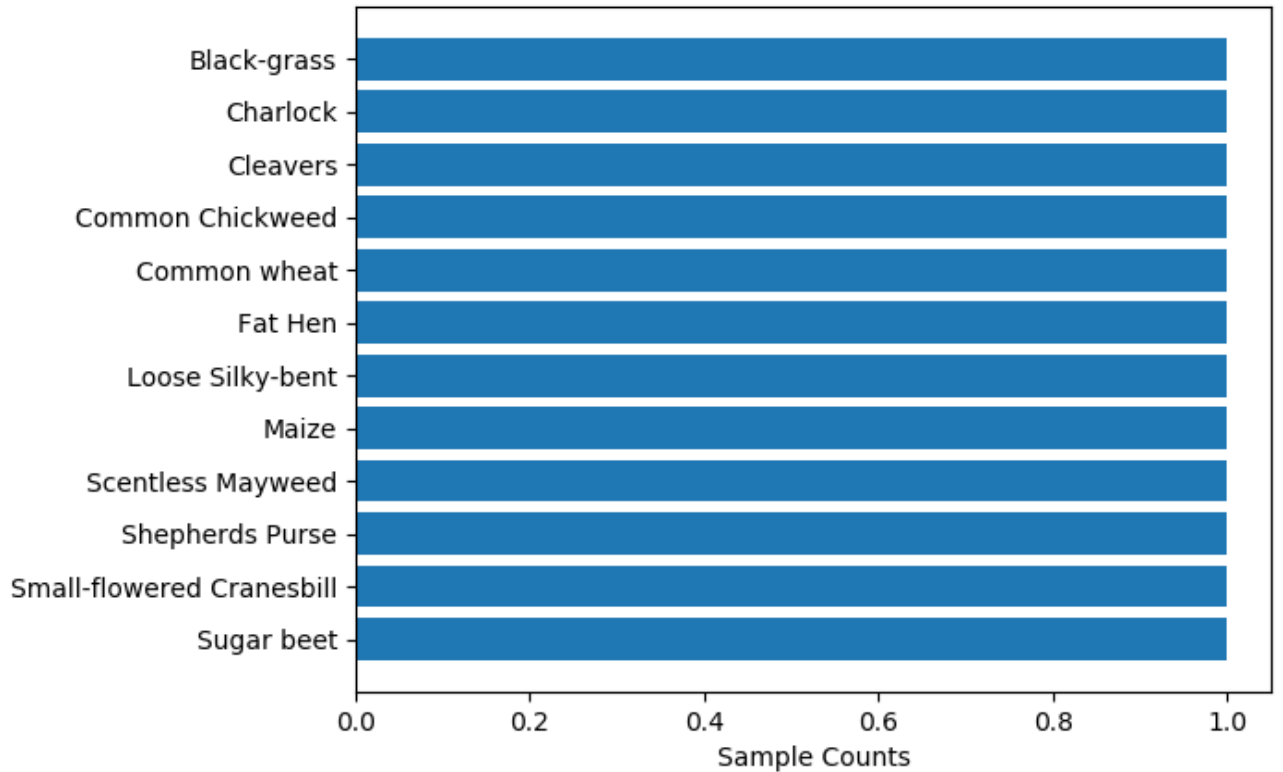
plt.show()

```

D:\seed classification model\train Sample Counts Per Class



D:\seed classification model\validation Sample Counts Per Class



```
In [7]: lower_bound= (24, 50, 0)
        upper_bound= (55, 255, 255)

fig= plt.figure(figsize=(10, 10))
fig.suptitle('Random Pre-Processed Image From Each Class', fontsize=14, y=.92,
horizontalalignment='center', weight='bold')

for i in range(12):
    sample_class=os.path.join(r'D:\seed Classification Model\test')
    random_image= os.path.join(sample_class, random.choice(os.listdir(sample_cl
ass)))
    img= cv2.imread(random_image)
    img= cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
    img= cv2.resize(img, (150, 150))

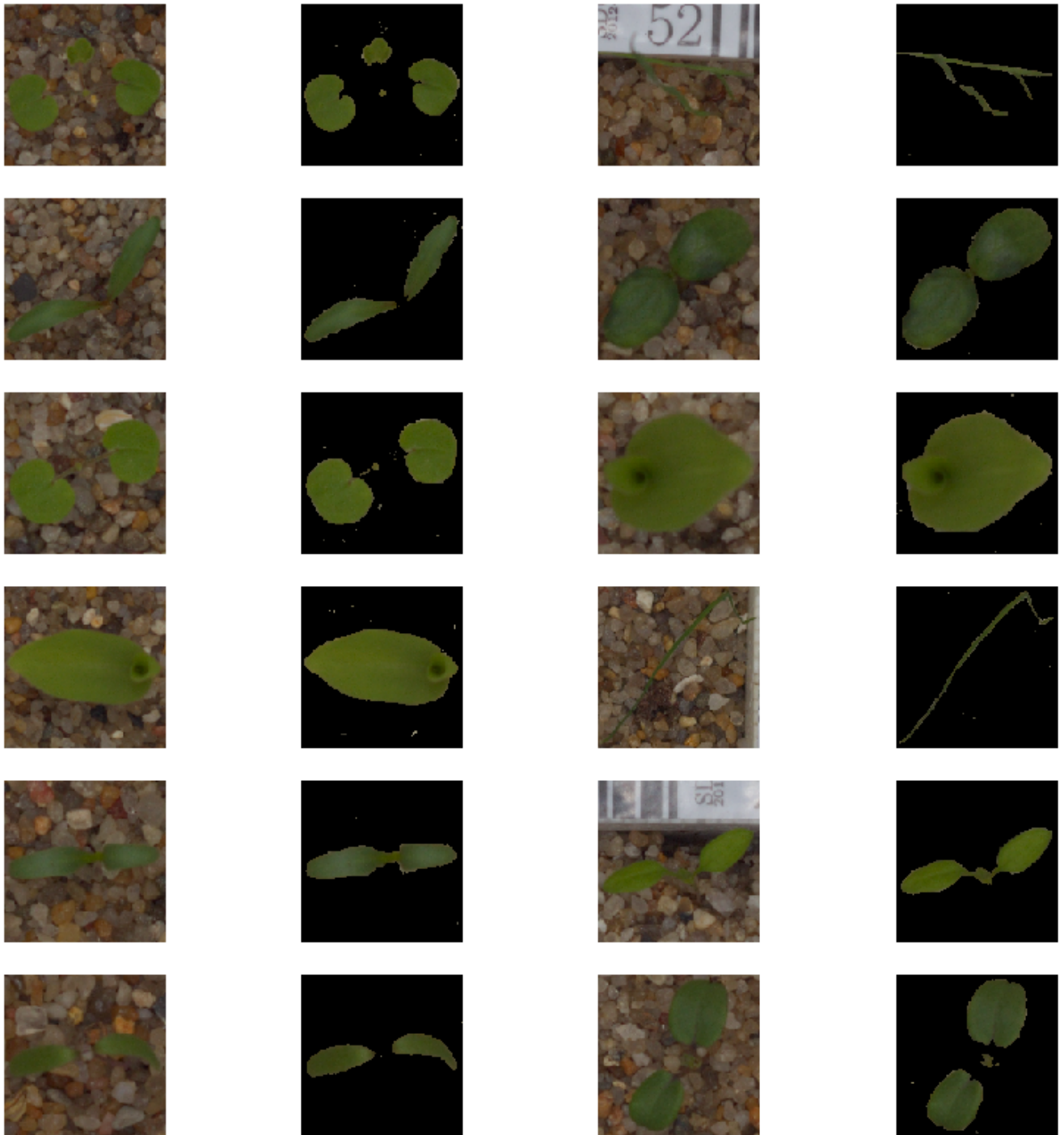
    hsv_img= cv2.cvtColor(img, cv2.COLOR_RGB2HSV)
    mask = cv2.inRange(hsv_img, lower_bound, upper_bound)
    result = cv2.bitwise_and(img, img, mask=mask)

    fig.add_subplot(6, 4, i*2+1)
    plt.imshow(img)
    plt.axis('off')

    fig.add_subplot(6, 4, i*2+2)
    plt.imshow(result)
    plt.axis('off')

plt.show()
```

Random Pre-Processed Image From Each Class



```
In [8]: def color_segment_function(img_array):  
        img_array= np.rint(img_array)  
        img_array= img_array.astype('uint8')  
        hsv_img= cv2.cvtColor(img_array, cv2.COLOR_RGB2HSV)  
        mask = cv2.inRange(hsv_img, (24, 50, 0), (55, 255, 255))  
        result = cv2.bitwise_and(img_array, img_array, mask=mask)  
        result= result.astype('float64')  
        return result
```

```
In [9]: #image function from keras.preprocessing
train_datagen = tf.keras.preprocessing.image.ImageDataGenerator(
    rotation_range=20,
    zoom_range=0.15,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.15,
    horizontal_flip=True,
    brightness_range=[0.4,1],
    rescale=1.0/255.0)

test_datagen = image.ImageDataGenerator(rescale=1./255, preprocessing_function=
color_segment_function)
```

```
In [10]: training_set = train_datagen.flow_from_directory(r'D:\seed Classification Model\train',
                                                         target_size = (150, 150),
                                                         batch_size = 20,
                                                         class_mode = 'categorical')
```

Found 4738 images belonging to 12 classes.

```
In [50]: test_set = test_datagen.flow_from_directory(r'D:\seed Classification Model\validation',
                                                      target_size = (150, 150),
                                                      batch_size = 20,
                                                      class_mode = 'categorical')
```

Found 12 images belonging to 12 classes.


```
In [12]: #Model Training part begins
model = models.Sequential()

model.add(layers.Conv2D(32, (3, 3), input_shape=(150, 150, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.1))

model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.1))

model.add(layers.Conv2D(128, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.1))

model.add(layers.Conv2D(128, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.1))

model.add(layers.Flatten())
model.add(layers.Dropout(0.4))
model.add(layers.Dense(256, activation='relu'))
model.add(layers.Dropout(0.4))

model.add(layers.Dense(12, activation='softmax'))
```

WARNING:tensorflow:From C:\Users\Parth Madan\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:4070: The name tf.nn.max_pool is deprecated. Please use tf.nn.max_pool2d instead.

```
In [13]: model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
=====		
conv2d_1 (Conv2D)	(None, 148, 148, 32)	896
max_pooling2d_1 (MaxPooling2D)	(None, 74, 74, 32)	0
dropout_1 (Dropout)	(None, 74, 74, 32)	0
conv2d_2 (Conv2D)	(None, 72, 72, 64)	18496
max_pooling2d_2 (MaxPooling2D)	(None, 36, 36, 64)	0
dropout_2 (Dropout)	(None, 36, 36, 64)	0
conv2d_3 (Conv2D)	(None, 34, 34, 128)	73856
max_pooling2d_3 (MaxPooling2D)	(None, 17, 17, 128)	0
dropout_3 (Dropout)	(None, 17, 17, 128)	0
conv2d_4 (Conv2D)	(None, 15, 15, 128)	147584
max_pooling2d_4 (MaxPooling2D)	(None, 7, 7, 128)	0
dropout_4 (Dropout)	(None, 7, 7, 128)	0
flatten_1 (Flatten)	(None, 6272)	0
dropout_5 (Dropout)	(None, 6272)	0
dense_1 (Dense)	(None, 256)	1605888
dropout_6 (Dropout)	(None, 256)	0
dense_2 (Dense)	(None, 12)	3084
=====		
Total params: 1,849,804		
Trainable params: 1,849,804		
Non-trainable params: 0		

```
In [14]: best_cb= callbacks.ModelCheckpoint('model_best.h5',
                                             monitor='val_loss',
                                             verbose=1,
                                             save_best_only=True,
                                             save_weights_only=False,
                                             mode='auto',
                                             period=1)

opt= keras.optimizers.Adam(lr=0.0005, amsgrad=True)  # Learning rate=0.00005
```

```
In [15]: model.compile(optimizer=opt,  
                        loss='categorical_crossentropy',  
                        metrics=['accuracy'])  
  
history = model.fit_generator(  
    training_set,  
    validation_data=test_set,  
    epochs=50,  
    steps_per_epoch=len(training_set),  
    validation_steps=len(test_set),  
    callbacks= [best_cb])
```

WARNING:tensorflow:From C:\Users\Parth Madan\Anaconda3\lib\site-packages\keras\backend\tensorflow_backend.py:422: The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

Epoch 1/50

237/237 [=====] - 334s 1s/step - loss: 2.2886 - accuracy: 0.2018

Epoch 2/50

C:\Users\Parth Madan\Anaconda3\lib\site-packages\keras\callbacks\callbacks.py:707: RuntimeWarning: Can save best model only with val_loss available, skipping.
'skipping.' % (self.monitor), RuntimeWarning)

237/237 [=====] - 242s 1s/step - loss: 1.8447 - accuracy: 0.3527
Epoch 3/50
237/237 [=====] - 243s 1s/step - loss: 1.5856 - accuracy: 0.4491
Epoch 4/50
237/237 [=====] - 385s 2s/step - loss: 1.4449 - accuracy: 0.5065
Epoch 5/50
237/237 [=====] - 347s 1s/step - loss: 1.2917 - accuracy: 0.5585
Epoch 6/50
237/237 [=====] - 260s 1s/step - loss: 1.2246 - accuracy: 0.5678
Epoch 7/50
237/237 [=====] - 231s 973ms/step - loss: 1.1328 - accuracy: 0.6173
Epoch 8/50
237/237 [=====] - 233s 982ms/step - loss: 1.0363 - accuracy: 0.6522
Epoch 9/50
237/237 [=====] - 230s 972ms/step - loss: 0.9924 - accuracy: 0.6638
Epoch 10/50
237/237 [=====] - 241s 1s/step - loss: 0.9020 - accuracy: 0.6933
Epoch 11/50
237/237 [=====] - 241s 1s/step - loss: 0.8682 - accuracy: 0.7075
Epoch 12/50
237/237 [=====] - 248s 1s/step - loss: 0.8252 - accuracy: 0.7227
Epoch 13/50
237/237 [=====] - 245s 1s/step - loss: 0.7613 - accuracy: 0.7419
Epoch 14/50
237/237 [=====] - 230s 973ms/step - loss: 0.7478 - accuracy: 0.7463
Epoch 15/50
237/237 [=====] - 232s 978ms/step - loss: 0.7096 - accuracy: 0.7539
Epoch 16/50
237/237 [=====] - 237s 998ms/step - loss: 0.6486 - accuracy: 0.7712
Epoch 17/50
237/237 [=====] - 245s 1s/step - loss: 0.6759 - accuracy: 0.7706
Epoch 18/50
237/237 [=====] - 232s 981ms/step - loss: 0.6352 - accuracy: 0.7860
Epoch 19/50
237/237 [=====] - 231s 976ms/step - loss: 0.5905 - accuracy: 0.7940
Epoch 20/50
237/237 [=====] - 231s 973ms/step - loss: 0.5658 - accuracy: 0.8022
Epoch 21/50
237/237 [=====] - 230s 970ms/step - loss: 0.5790 - accuracy: 0.8008
Epoch 22/50
237/237 [=====] - 246s 1s/step - loss: 0.5469 - accuracy:

cy: 0.8128
Epoch 23/50
237/237 [=====] - 230s 971ms/step - loss: 0.5344 - accuracy: 0.8134
Epoch 24/50
237/237 [=====] - 229s 966ms/step - loss: 0.5186 - accuracy: 0.8179
Epoch 25/50
237/237 [=====] - 327s 1s/step - loss: 0.5019 - accuracy: 0.8219
Epoch 26/50
237/237 [=====] - 243s 1s/step - loss: 0.4969 - accuracy: 0.8246
Epoch 27/50
237/237 [=====] - 270s 1s/step - loss: 0.4824 - accuracy: 0.8316
Epoch 28/50
237/237 [=====] - 257s 1s/step - loss: 0.4507 - accuracy: 0.8404
Epoch 29/50
237/237 [=====] - 270s 1s/step - loss: 0.4535 - accuracy: 0.8438
Epoch 30/50
237/237 [=====] - 248s 1s/step - loss: 0.4600 - accuracy: 0.8371
Epoch 31/50
237/237 [=====] - 279s 1s/step - loss: 0.4598 - accuracy: 0.8364
Epoch 32/50
237/237 [=====] - 304s 1s/step - loss: 0.4414 - accuracy: 0.8407
Epoch 33/50
237/237 [=====] - 257s 1s/step - loss: 0.4192 - accuracy: 0.8607
Epoch 34/50
237/237 [=====] - 249s 1s/step - loss: 0.4187 - accuracy: 0.8520
Epoch 35/50
237/237 [=====] - 278s 1s/step - loss: 0.3709 - accuracy: 0.8645
Epoch 36/50
237/237 [=====] - 499s 2s/step - loss: 0.3842 - accuracy: 0.8594
Epoch 37/50
237/237 [=====] - 401s 2s/step - loss: 0.4021 - accuracy: 0.8573
Epoch 38/50
237/237 [=====] - 360s 2s/step - loss: 0.3777 - accuracy: 0.8628
Epoch 39/50
237/237 [=====] - 419s 2s/step - loss: 0.3738 - accuracy: 0.8691
Epoch 40/50
237/237 [=====] - 401s 2s/step - loss: 0.3775 - accuracy: 0.8628
Epoch 41/50
237/237 [=====] - 597s 3s/step - loss: 0.3747 - accuracy: 0.8687
Epoch 42/50
237/237 [=====] - 360s 2s/step - loss: 0.3552 - accuracy: 0.8725
Epoch 43/50

```
237/237 [=====] - 306s 1s/step - loss: 0.3566 - accuracy: 0.8736
Epoch 44/50
237/237 [=====] - 325s 1s/step - loss: 0.3619 - accuracy: 0.8681
Epoch 45/50
237/237 [=====] - 302s 1s/step - loss: 0.3609 - accuracy: 0.8681
Epoch 46/50
237/237 [=====] - 257s 1s/step - loss: 0.3539 - accuracy: 0.8721
Epoch 47/50
237/237 [=====] - 242s 1s/step - loss: 0.3397 - accuracy: 0.8808
Epoch 48/50
237/237 [=====] - 282s 1s/step - loss: 0.3279 - accuracy: 0.8837
Epoch 49/50
237/237 [=====] - 252s 1s/step - loss: 0.3392 - accuracy: 0.8797
Epoch 50/50
237/237 [=====] - 299s 1s/step - loss: 0.3230 - accuracy: 0.8850
```

```
In [27]: from tensorflow.keras.models import load_model
        model.save('model_best.h5')
```

```
In [ ]:
```

```
In [28]: #load best model from training
        model= models.load_model('model_best.h5')
```

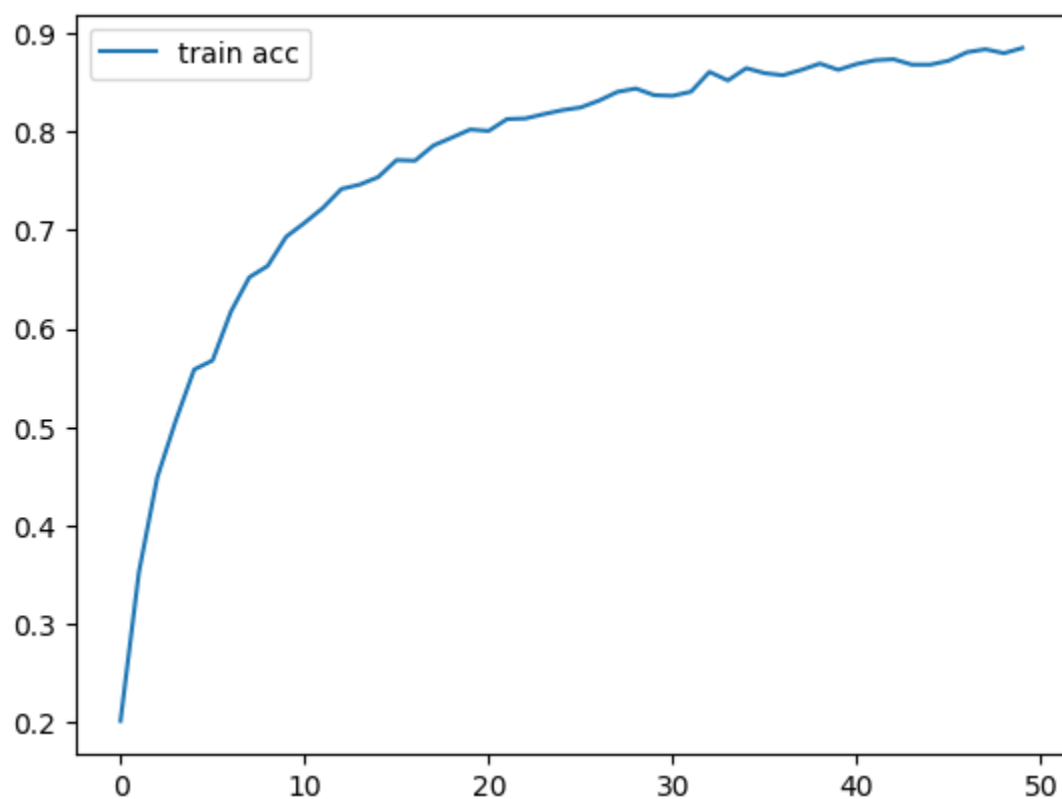
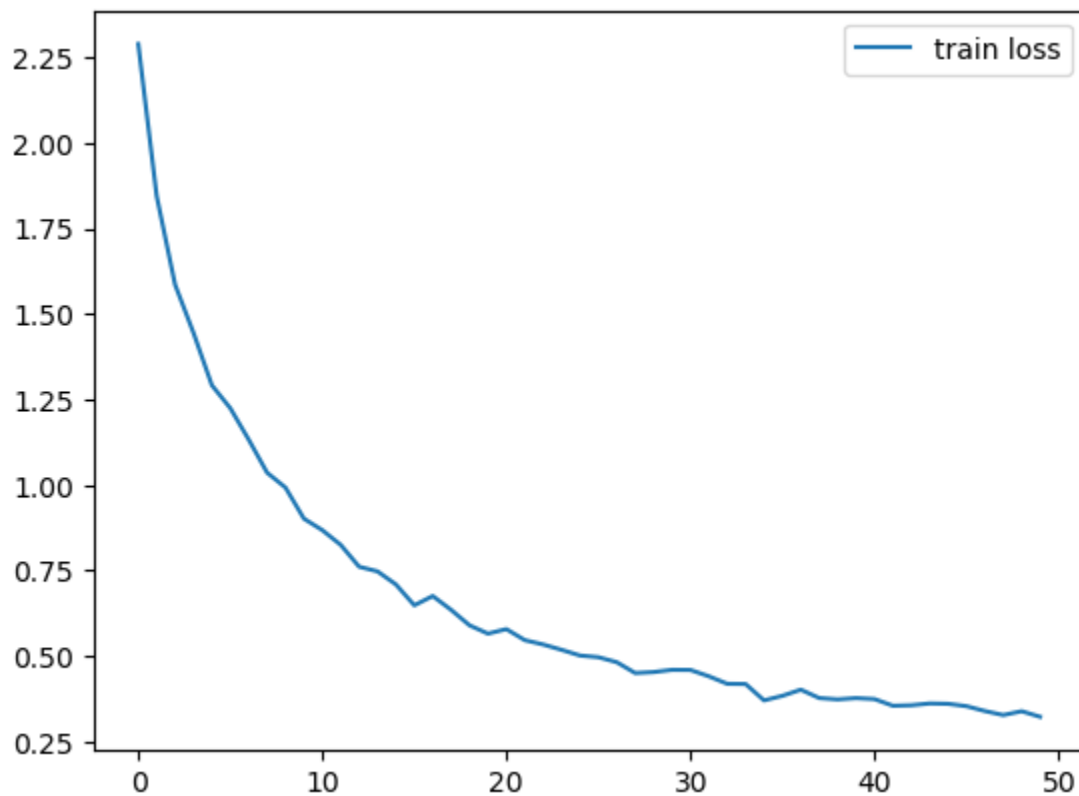
```
In [51]: model
```

```
Out[51]: <keras.engine.sequential.Sequential at 0x191801a8128>
```

```
In [29]: #save history
        with open('model_history.pkl', 'wb') as f:
            pickle.dump(history, f)
```

```
In [46]: # plot the loss
import matplotlib.pyplot as plt
plt.plot(history.history['loss'], label='train loss')
#plt.plot(history.history['val_loss'], label='val loss')
plt.legend()
plt.show()
plt.savefig('LossVal_loss')

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

<Figure size 640x480 with 0 Axes>

```
In [53]: pred= model.predict_generator(test_set, steps= test_set.n, verbose=1)
12/12 [=====] - 7s 610ms/step
```

```
In [70]: from utils import label_map_util
predicted_class_indices=np.argmax(pred,axis=1)

prediction_labels = [label_map_util[k] for k in predicted_class_indices]

-----
ImportError                                Traceback (most recent call last)
<ipython-input-70-0cf1c08cab7b> in <module>
----> 1 from utils import label_map_util
      2 predicted_class_indices=np.argmax(pred,axis=1)
      3
      4 prediction_labels = [label_map_util[k] for k in predicted_class_indices
      ]

ImportError: cannot import name 'label_map_util' from 'utils' (C:\Users\Parth M
adan\Anaconda3\lib\site-packages\utils\__init__.py)
```

```
In [63]: filenames= test_set.filenames
```

```
In [64]: #Final Result time for predicting the soln for some images dataset To check our
model valid or not for predicting the results.
import csv
csvfile= open(r'D:\seed Classification Model\result_sample', 'w', newline='')
writer= csv.writer(csvfile)
headers= ['file', 'species']

writer.writerow(headers)
t = PrettyTable(headers)
for i, f, p in zip(range(len(filenames)), filenames, prediction_labels):
    writer.writerow([os.path.basename(f),p])
    if i <10:
        t.add_row([os.path.basename(f), p])
    elif i<13:
        t.add_row(['.', '.'])
csvfile.close()
print(t)
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-64-10aa769677c6> in <module>
      7 writer.writerow(headers)
      8 t = PrettyTable(headers)
----> 9 for i, f, p in zip(range(len(filenames)), filenames, prediction_labels)
      :
     10     writer.writerow([os.path.basename(f),p])
     11     if i <10:

NameError: name 'prediction_labels' is not defined
```

```
In [42]:
```

```
Out[42]: []
```

```
In [ ]:
```