

Plant Disease Classification

(using leaf images)

Submitted By:-

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Under the guidance of :-

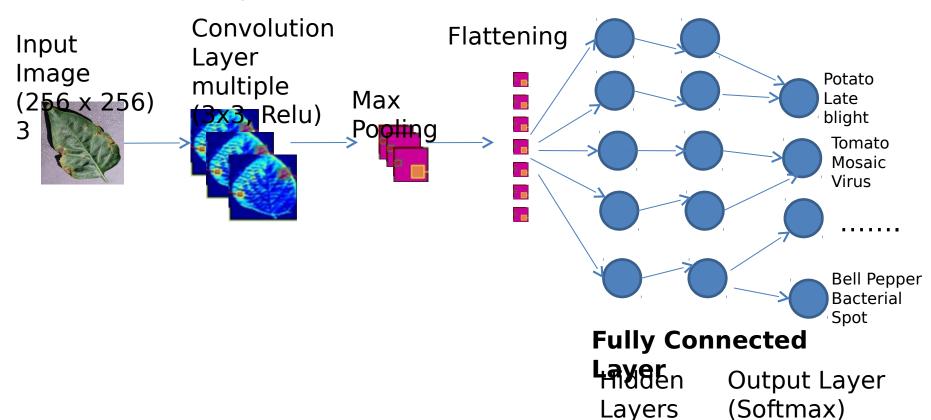
Mrs. Vimala Mathew

14th January 2020



Solution Design

Objective: To classify leaf images (healthy and disease) using CNN





Training Approach

- •Model is trained on around **20,000 leaf images** (10 Epochs and 32 as Batch size)
- Images split into Training and Test
- Image Augmentation applied for each image
- Performance metrics are plotted (accuracy and loss)
- Trained model saved and used to classify a leaf image (healthy or disease)

Program: Train model for Classificat

```
#Import neccessary packages
from os import listdir
import sys
import numpy as np
import cv2
import matplotlib.pyplot as plt
from keras.models import Sequential
from keras.layers.convolutional import Conv2D, MaxPooling2D
from keras.layers.core import Activation, Flatten, Dense
from keras import backend as K
from keras.optimizers import Adam
from keras.preprocessing import image
from keras.preprocessing.image import ImageDataGenerator,img to array
from keras.utils import to categorical
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import LabelBinarizer
from sklearn.model selection import train test split
#Initialise few vars
EPOCHS = 10
INIT LR = 1e-3
BS = 32
default image size = tuple((256, 256))
image size = 0
directory root = '/home/ai16/project/main project/Data/Plantvillage'
width=256
height=256
depth=3
```



```
#Function to convert images to array
def convert image to array(image dir):
    try:
        image = cv2.imread(image dir)
        if image is not None :
            image = cv2.resize(image, default image size)
            return img to array(image)
        else :
            return np.array([])
    except Exception as e:
        print("convert image to array() : Error when converting Image to Array:"+e)
        return None
#Fetch images from directory and convert each to an array...
#assign labels to it (folder name)
image list, label list = [], []
try:
   print("[INFO] Loading images from folders...")
    root dir = listdir(directory root)
    for plant folder in root dir :
        print("Processing images from folder... ->: "+ plant folder)
        Images In Folder = listdir(directory root+"/"+plant folder)
        #for image in plant disease folder list[:200]:
        for image in Images In Folder:
                image filename = directory root+"/"+plant folder+"/"+image
                if image_filename.endswith(".jpg") == True or image filename.endswith(".JPG")
                    image list.append(convert image to array(image filename))
                    label list.append(plant folder)
```



```
print("[INFO] Image loading completed from all directories....")
except Exception as e:
    print("Try...Catch: Error when loding/processing images...: "+str(e))
#Convert labels to numeric values (Ex. 0,1,2,3...based on categories)
le=LabelEncoder()
label list num=le.fit transform(label list)
label classes = le.classes
n classes=len(np.unique(label list num))
#convert to binary values (one hot encoding)
label list num bin=to_categorical(label_list_num)
#Tensor Flow compatible (4-d array) and Normalize the pixels
np image list = np.array(image list, dtype=np.float16) / 255.0
#Create Train and Test set
print("[INFO] Spliting data to train, test")
x train, x test, y train, y test = train test split(np image list, label list num bin, test s
#Image Augmentation
aug = ImageDataGenerator(
    rotation range=25,
    width shift range=0.1,
    height shift range=0.1,
    shear range=0.2,
    zoom range=0.2,
    horizontal flip=True,
    fill mode="nearest")
```



```
inputShape = (height, width, depth)
#Build CNN layers
model = Sequential()
model.add(Conv2D(32, (3, 3), input shape=inputShape))
model.add(Activation("relu"))
model.add(MaxPooling2D(pool size=(3, 3)))
model.add(Conv2D(64, (3, 3)))
model.add(Activation("relu"))
model.add(Conv2D(64, (3, 3)))
model.add(Activation("relu"))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Conv2D(128, (3, 3)))
model.add(Activation("relu"))
model.add(Conv2D(128, (3, 3)))
model.add(Activation("relu"))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Flatten())
```



```
#Fully Connected Layer
model.add(Dense(128))
model.add(Activation("relu"))
#Output layer
model.add(Dense(n classes))
model.add(Activation("softmax"))
#model.summary()
opt = Adam(lr=INIT LR)
#Complile the model
model.compile(loss="categorical crossentropy", optimizer=opt,metrics=["accuracy"])
# train the network
print("[INFO] training network...")
history = model.fit generator(
    aug.flow(x train, y train, batch size=BS),
    validation data=(x test, y test),
    steps per epoch=len(x train) // BS,
    epochs=EPOCHS, verbose=1
acc = history.history['acc']
val acc = history.history['val_acc']
loss = history.history['loss']
val loss = history.history['val loss']
epochs = range(1, len(acc) + 1)
```



```
acc = history.history['acc']
val acc = history.history['val acc']
loss = history.history['loss']
val loss = history.history['val loss']
epochs = range(1, len(acc) + 1)
#Plot Accuracy and Loss for Training and Validation set
#Accuracy
plt.plot(epochs, acc, 'b', label='Training accurarcy')
plt.plot(epochs, val acc, 'r', label='Validation accurarcy')
plt.title('Training and Validation accurarcy')
plt.legend()
plt.figure()
#Loss
plt.plot(epochs, loss, 'b', label='Training loss')
plt.plot(epochs, val loss, 'r', label='Validation loss')
plt.title('Training and Validation loss')
plt.legend()
plt.show()
#Model Accuracy
print("[INFO] Calculating model accuracy")
scores = model.evaluate(x test, y test)
print("Test Loss: "+str(scores[0]))
print("Test Accuracy: "+str(scores[1]*100))
#Save the model for prediction
print("[INFO] Saving model...")
model.save("cnn model.h5")
```

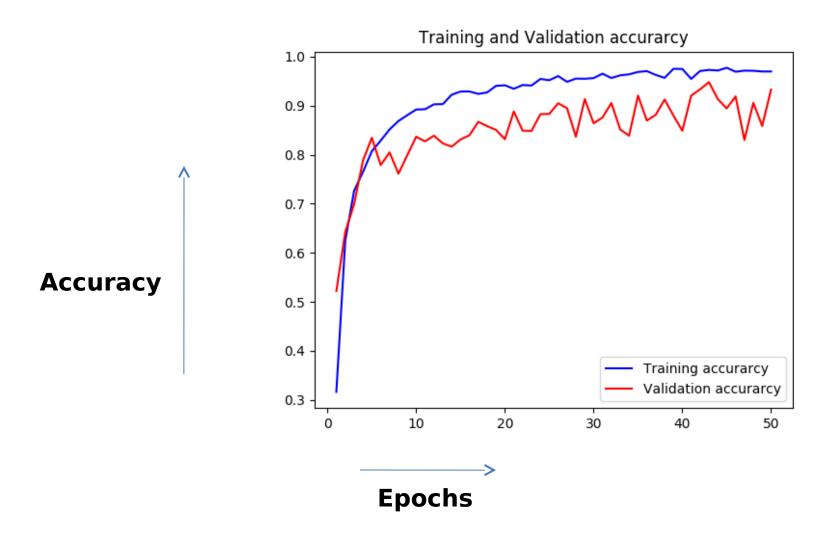
Program: Image classification based-Trained model

```
from keras.preprocessing import image
from keras.models import load model
import matplotlib.pyplot as plt
import numpy as np
import os
import cv2 as cv
def load image (img path, show=False):
    img = image.load img(img path, target size=(256, 256))
    img array = image.img to array(img)
    img array = np.expand dims(img array, axis=0)
    if show:
       plt.imshow(img array[0])
       plt.axis('off')
       plt.show()
    return img array
if name == " main ":
    #load model
   model = load model("/home/ai16/project/main project/cnn model.h5")
    #image path
   img_path = '/home/ai16/project/main_project/Data/TEST/Potato_Early_blight/0a8a68ee-f587-4
    #Load a single image
   new image = load image(img path)
```



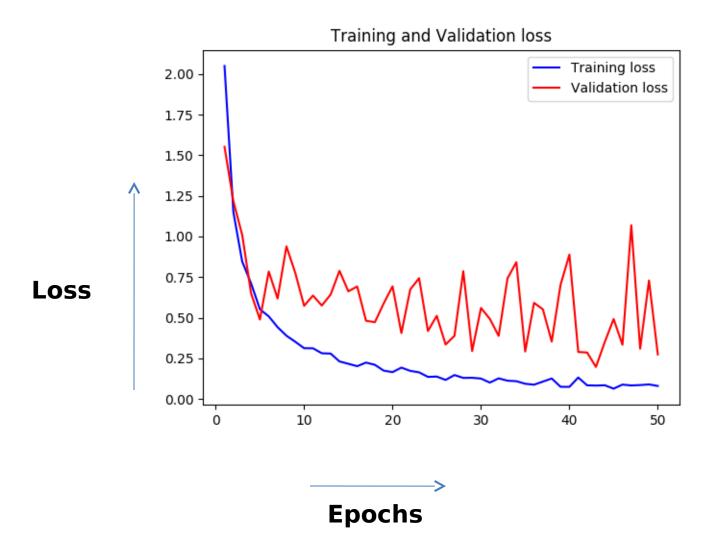
```
if name == " main ":
    #load model
   model = load model("/home/ai16/project/main project/cnn model.h5")
    #image path
    img path = '/home/ai16/project/main project/Data/TEST/Potato Early blight/0a8a68ee-f587-4
    #Load a single image
   new image = load image(img path)
    #Availbale leaf image categories
   new list = ['Pepper bell Bacterial spot', 'Pepper bell healthy', 'Potato Early blight',\
 'Potato Late blight', 'Potato healthy', 'Tomato Bacterial spot',\
 'Tomato Early blight', 'Tomato Late blight', 'Tomato Leaf Mold',\
 'Tomato Septoria leaf spot', 'Tomato Spider mites', 'Tomato Target Spot',\
 'Tomato YellowLeaf Curl Virus', 'Tomato healthy', 'Tomato mosaic virus']
    #Predict the image
    print("Probability of the image across categories.....")
    pred = model.predict(new image)
    print("Image classified as....")
    classes names=pred.argmax(axis=-1)
    #print(classes names)
   print(new list[classes names[0]])
    img vw=cv.imread(img path)
    cv.imshow('Input Image', img vw)
    cv.waitKey(10000)
```

Training & Validation Accuracy (12,000 leaf images)





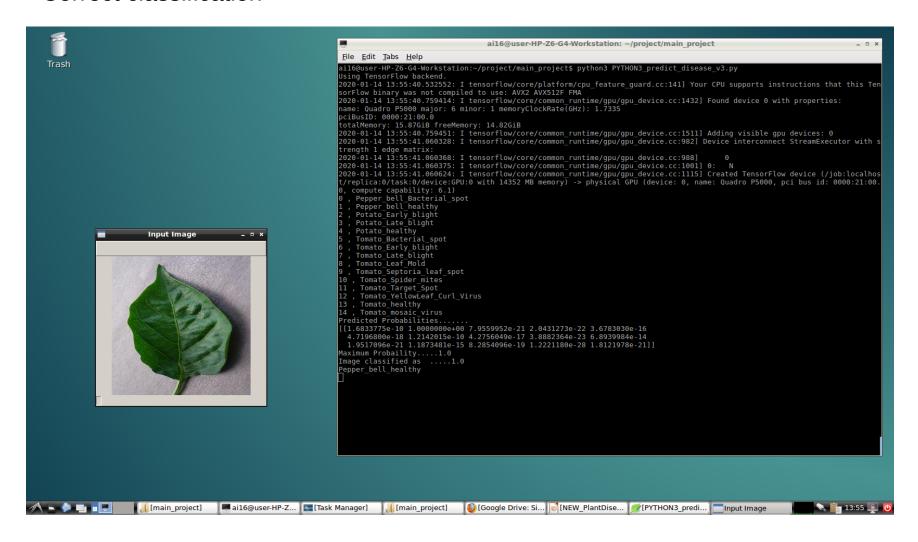
Training & Validation Loss (12,000 leaf images)





Output

Correct classification





Future Scope

- Highlight the infected leaf part
- Increase coverage of plants for disease identification
- Remedy for a plant disease can be provided
- Model can be integrated with a mobile app and farmers can use (with local language support)



Conclusion

 CNN model was able to classify the leaf images as healthy or disease with probability

 Training accuracy improved and loss decreased with increase in number of samples (from 1500 to 20,000)



Thank You