

Experiment No. 3 : Plant Disease prediction using CNN (PART 1)

```
import numpy as np

import pickle

import cv2

from os import listdir

from sklearn.preprocessing import LabelBinarizer

from keras.models import Sequential

from keras.layers.normalization import Batch Normalization

from keras.layers.convolutional import Conv2D

from keras.layers.convolutional import MaxPooling2D

from keras.layers.core import Activation, Flatten, Dropout, Dense

from keras import backend as K

from keras.preprocessing.image import ImageDataGenerator

from keras.optimizers import Adam

from keras.preprocessing import image

from keras.preprocessing.image import img_to_array

from sklearn.preprocessing import MultiLabelBinarizer

from sklearn.model_selection import train_test_split

import matplotlib.pyplot as plt

import tensorflow

EPOCHS = 25

INIT_LR= 1e-3

BS = 32

default image_size = tuple((256, 256))

image_size = 0
```

```

directory_root = './input/plantvillage/'

width=256

height=256

depth=3

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(f"Error : {e}")

    return None

image_list, label_list = [], []

try:

    print("[INFO] Loading images..")

    root_dir = listdir (directory_root)

    for directory in root_dir :

        # remove .DS_Store from list

        if directory == ".DS_Store":

            root_dir.remove (directory)

```

```

for plant_folder in root_dir :

plant_disease_folder_list = listdir(f" {directory_root}/{(plant_folder)}")


for disease_folder in plant_disease_folder list:

# remove .DS_Store from list

if disease folder == ".DS_Store" :

plant_disease _folder_list.remove (disease_folder)


for plant_disease_folder in plant_disease_folder_list:

print(f"[INFO] Processing {plant disease_folder}..")

plant_disease_image_list = listdir(f"{directory_root}/{plant_folder}/{plant_disease_folder} /")


for single_plant_disease_image in plant_disease_image_list :

if single_plant_disease_image == ".DS_ Store":

plant_disease_image_list.remove(single_plant_disease_image)


for image in plant_disease_image_list[:200]:

image_directory = f"{directory_root}/{plant, folder}/{plant_disease_folder}/{image}"

if image_directory.endswith (".jpg") == True or image_directory.endswith(".JPG") == True:

Image_list.append (convert_image_to_array(image_directory))

label_list.append (plant_disease_folder)

print("[INFO] Image loading completed")

except Exception as e:

print(f"Error:{e}")

image_size = len(image_list)

```

```

label_binarizer = LabelBinarizer()

image_labels = label_binarizer.fit_transform(label_list)

pickle.dump(label_binarizer,open('label_transform.pkl', 'wb'))

n_classes = len(label_binarizer.classes_)


print(label_binarizer.classes)


np_image_list = np.array (image_list, dtype=np.float16) / 225.0


print("[INFO] Splitting data to train, test")

X_train, x_test, y_train, y_test = train_test_split(np_image_list, image_labels, test_size=0.2,
random state = 42)


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")


model = Sequential()

inputShape = (height, width, depth)

    chanDim=-1

if K.image_data_format() == "channels_first":

inputShape = (depth, height, width)

    chan Dim=1

model.add (Conv2D(32, (3, 3), padding="same",input_shape=inputShape))

model.add(Activation("relu"))

```

```
model.add (Batch Normalization(axis=chanDim))
model.add (MaxPooling2D(pool_size=(3, 3)))
model.add (Dropout(0.25))
model.add (Conv2D (64, (3, 3), padding="same"))
model.add (Activation("relu"))
model.add(BatchNormalization(axis=chanDim))
model.add(Conv2D(64,(3, 3),padding="same"))

model.add(Activation("relu")

model.add (Batch Normalization(axis=chanDim))
model.add(MaxPooling2D(poo_size=(2, 2)))
model.add(Dropout(0.25))
model.add(Conv2D(128, (3, 3),. padding="same"))
model.add(Activation("relu"))

model.add(BatchNormalization(axis=chan Dim))
model.add(Conv2D (128, (3, 3),padding="same"))
model.add(Activation("relu"))

model.add(BatchNormalization(axis=chanDim))
model.add(MaxPooling2D(poolsize=(2, 2))
model.add(Dropout(0.25))

model.add(Flatten())
model.add(Dense(1024))
model.add(Activation("relu"))
model.add(BatchNormalization())
model.add(Dropout(0.5))
model.add (Dense(n_classes))
model.add (Activation("softmax"))
```

```
opt= Adam(lr=INIT_LR, decay=INIT_LR / EPOCHS)

# distribution

model.compile(loss="binary_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)
```

```
#Train and validation accuracy

plt.plot(epochs, acc, 'b', label='Training accuracy')

plt.plot(epochs, val_acc, 'r', label='Validation accuracy')

plt.title('Training and Validation accuracy')

plt.legend()

plt.figure()
```

```
#Train and validation 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()
```

```
print("[INFO] Calculating model accuracy")
scores = model.evaluate(x_test, y_test)
print(f"Test Accuracy: {scores[1]*100}")
```

```
# save the model to disk
print("[INFO] Saving model..")
pickle.dump(model,open('cnn_model.pkl', 'wb'))

Accuracy: 96.77%
[INFO] Calculating model accuracy
591/591[=====] - 2s 3ms/step
Test Accuracy : 96.773830807551 92
```

Experiment No. 3 : Classification of MNIST Fashion dataset using CNN(PART 2)

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from keras.utils import to_categorical
from matplotlib.pyplot import figure, show
import warnings
import seaborn as sns
warnings.filterwarnings(ignore)
import matplotlib.style as style
```

```
from sklearn.model_selection import train_test_split

from keras.layers import Input, Concatenate, concatenate, Dense, Embedding, Dropout,
Conv2D, MaxPooling2D

from keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau, History

from keras.layers import Dropout, Flatten, GlobalAveragePooling2D, Activation

from sklearn.preprocessing import LabelEncoder, OneHotEncoder

from keras.preprocessing.image import imageDataGenerator

from sklearn.model_selection import train_test_split

from keras.applications.resnet50 import ResNet50

from keras.callbacks import ReduceLROnPlateau

from keras.callbacks import ModelCheckpoint

from keras.applications.vgg16 import VGG16

from keras.utils import to_categorical

from sklearn.utils import class_weight

from keras.layers.normalization import BatchNormalization

from matplotlib import pyplot as plt

from keras import backend as K

from keras.optimizers import SGD

from keras.models import Model

import seaborn as sns

import numpy as np

import argparse

import time

import glob

import cv2

import numpy

import os
```



```
import glob
import sys
import os
import json
import pprint
import warnings
warnings.filterwarnings('ignore')
```

```
#Load Data
```

```
!curl -L -O https://www.dropbox.com/s/heyql2my8uwotq/fashionmaistzip
```

```
!unzip fashionmnist.zip
```

```
#Load training and test data using dataframes from Pandas.
```

```
train = pd.read_csv("fashion-mnist_train.csv")
```

```
test = pd.read_csv("fashion-mnist_test.csv")
```

```
img_rows, img_cols = 28, 28
```

```
input_shape = (img_rows, img_cols, 1)
```

```
X= train.iloc[:,1:]
```

```
Y= train.iloc[:,0]
```

```
X_test = test.iloc[:, 1:]
```

```
Y_test = test.iloc[:, 0]
```

```
#Normalization
```

```
X= np.asarray(X).reshape (X.shape [0], img_rows,img_cols, 1)
```

```
X_test = np.asarray(X_test).reshape(X_test.shape [0], img_rows,img_cols,1)
```

$X = (255. - X) / 255.$

$X_test = (255. - X_test) / 255.$

#Number of classes

classes = len(Y['label'],value _counts())

print("Number of features: ", X.shape[1])

print("Number of train samples: ", Xshape [0])

print("Number of test samples: ", X test.shape [0])

OUT:

Number of features: 28

Number of train samples: 60000

Number of test samples: 10000

#Training

Y_test = to_categorical(Y_test)

Y= to_categorical (Y)

X_train, X_val, Y_train, Y_val = train_test_split(X, Y, stratify=Y, test_size=0.2,
random_state=66)

Irr = ReduceLROn Plateau (monitor='val_loss', factor=0.1, patience=2, verbose=1,
epsilon=1e-3, mode= 'min')

early_stopping = EarlyStopping(monitor='val loss',patience=5,verbose=0, mode='auto')

checkpoint = ModelCheckpoint("checkpoint.hdf5", monitor='val_acc', verbose=1,
save_best_only=True, mode='max')

batch size = 64

epochs = 10

```

from sklearn.model_selection import GridSearchCV

from keras.wrappers.scikit_learn import KerasClassifier


# define the grid search parameters

batch_size = [16, 32, 64, 80]

epochs = [10, 25, 50]

param_grid = dict (batch_size=batch_size, epochs=epochs)

model = KerasClassifier(build_fn=model_basic, verbose=0)

Grid=GridSearchCV(estimator=model, param_grid=param_grid,n_jobs1, cv=3)

grid_result = grid.fit(X_train, Y_train)


# summarize results

print("Best: %using %s" % (grid_result.best_score, grid_result.bestparams_))

means = grid_result.cv_results_['mean_test_score']

stds = grid_result.cv_results_['std_test_score']

params =grid_result.cv_results_['params']

for mean, stdev, param in zip(means, stds, params):

print("%f (%f) with: %r" % (mean, stdev, param))


#Model

def model_basic(classes=classes,optimizer='adam'):

kernel_size = (3,3)

dropout = 0.25

pool_size = (2,2)

inputs = Input(shape=(img_rows, img_cols, 1))

```

```
y= Conv2D (filters=32, kernel_size=kernel_size,activation='relu',padding='same') (inputs)
```

```
y= MaxPooling2D(pool_size=pool_size,strides=(2,2)) (y)
```

```
y= Dropout(dropout)(y)
```

```
y= Flatten()(y)
```

```
y= Dense(256,activation='relu')(y)
```

```
y= BatchNormalization()(y)
```

```
y= Dropout(dropout)(y)
```

```
outputs = Dense(classes, activation='softmax')(y)
```

```
model = Model(inputs=inputs, outputs=outputs)
```

```
model.compile (optimizer=optimizer, loss='categorical_crossentropy', metrics=['accuracy'])
```

```
return model
```

```
basic_model = model_basic()
```

```
import warnings
```

```
warnings.filterwarnings('ignore')
```

```
history = basic_model.fit(X_train, Y_train, batch_size=batch_size, epochs=epochs,
```

```
verbose=1, validation_data=(X_val, Y_val)
```

OUT:

Train on 48000 samples, validate on 12000 samples

Epoch 1/10

48000/48000 [=====]-82s Zms/step- loss: 0.4390 - acc:.8470 - val_loss: 0.4019val _acc:

0.8618

Epoch 2/10

48000/48000 [=====]-82s 2ms/step - loss: 0.3458 - acc: 0.8776 - val_loss: 0.3046
val_acc: 0.8932

Epoch 3/10

48000/48000 [=====] -82s 2ms/step - loss: 0.3110 - acc: 0.8883 - val_loss: 0.2947 -
val_acc: 0.8953

Epoch 4/10

48000/48000 [===]-81s 2ms/step - loss: 0.2935 - acc: 0.8937 - val_loss: 0.2772 -
val_acc: 0.9024

Epoch 5/10

48000/48000 [=====]-85s 2ms/step- loss: 0.2710 - acc: 0.9030 - val_loss: 0.2855-
val_acc: 0.8952

Epoch 6/10[=====| -84s 2ms/step - loss: 0.2592 - acc: 0.9067 - val _loss: 0.2574-
val_acc: 0.9063

Epoch 7/10

48000/48000 [=====]-85s 2ms/step-loss: 0.2450 - acc: 0.9107 - val_loss: 0.2773 -
val acc: 0.8998

Epoch 8/10

48000/48000 [=====] - 84s 2ms/step - loss: 0.2338 - acc: 0.9147 – val_loss: 0.2833 -
val_acc: 0.8955

Epoch 9/10

48000/48000 [=====]-82s 2ms/step - loss: 0.2239 - acc: 0.9174 – val_loss: 0.2553 -
val_acc: 0.9127

Epoch 10/10

48000/48000 [=====]-84s 2ms/step - loss: 0.2153 - acc: 0.9207 - val loss: 0.2564 -
val_acc: 0.9123

score = basic_model.evaluate(X_test,Y_test, verbose=0)

```
print("Test loss:", score[0])
```

```
print("Test accuracy:", score[1])
```

OUT:

Test loss: 0.2463475521683693

Test accuracy: 0.9137