#### **SPRINT 2**

Date : 07.11.2022

**Team ID** : PNT2022TMID22667

Project Name : FERTILIZERS

**RECOMMENDATION SYSTEM** 

FOR DISEASE PREDICTION

### **IMPORT THE LIBRARIES**

from keras.models import Sequential

from keras.layers import Dense

from keras.layers import Convolution2D

from keras.layers import MaxPooling2D

from keras.layers import Flatten

#### INITIALIZING THE MODEL

Keras has 2 ways to define a neural network:

- Sequential
- Function API

The Sequential class is used to define linear initializations of network layers which then, collectively, constitute a model.

We will use the Sequential constructor to create a model, which will then have layers added to it using the add ()method. Now, will initialize our model. Initialize the neural network layer by creating a reference/object to the Sequential class.

model = Sequential()

## **ADD CNN Layers**

## We will be adding three layers for CNN

- Convolution layer
- Pooling layer
- Flattening layer

```
model.add(Convolution2D(32,(3,3),input\_shape = (128,128,3),activation = 'relu')) \\ model.add(MaxPooling2D(pool\_size = (2,2))) \\ model.add(Flatten())
```

### ADD DENSE LAYERS

This step is to add a dense layer (output layer) where you will be specifying the number of classes your dependent variable has, activation function, and weight initializer as the arguments. We use the add () method toadd dense layers. the output dimensions here is 6

```
model.add(Dense(40, 'relu'))
model.add(Dense(20, 'relu'))
model.add(Dense(6, 'softmax', ))
```

## TRAIN AND SAVE THE MODEL

#### **COMPILE THE MODEL**

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

# **Model.Summary()**

Can be used to see all parameters and shapes in each layer in our models.

model.summary()

# FIT AND SAVE THE MODEL

model.fit(x\_train,epochs=20,steps\_per\_epoch=89,validation\_data = x\_test, validation\_steps = 27)

The weights are to be saved for future use. The weights are saved in as .h5 file using save().

model.save("fruit.h5")

## **OUTPUT:**

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 126, 126, 32)	896
max_pooling2d ( MaxPooling2D )	(None, 63, 63, 32)	0
flatten (Flatten)	(None, 127008)	0

dense (Dense)	(None, 300)	38102700
dense_1 (Dense)	(None, 150)	45150
dense_2 (Dense)	(None, 75)	11325
dense_3 (Dense)	(None, 9)	684
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Total params: 38,160,755

Trainable params: 38,160,755

Non-trainable params: 0

# Epoch 1/20

89/89 [=======] - 52s 576ms/step - loss: 2.4956 - accuracy: 0.2686 - val\_loss: 246.3766 - val\_accuracy:

0.3426

Epoch 2/20

89/89 [=======] - 44s 498ms/step -

loss: 1.2983 - accuracy: 0.5468 - val\_loss: 651.4410 - val\_accuracy:

0.2894

**Epoch 3/20** 

```
89/89 [======] - 42s 469ms/step -
loss: 0.9406 - accuracy: 0.6735 - val_loss: 1125.0737 - val_accuracy: 0.2442
Epoch 4/20
89/89 [======] - 39s 440ms/step -
loss: 0.7779 - accuracy: 0.7300 - val_loss: 1022.7507 - val_accuracy:0.2847
Epoch 5/20
89/89 [=======] - 41s 462ms/step -
loss: 0.7470 - accuracy: 0.7465 - val loss: 1396.1002 - val accuracy: 0.2581
Epoch 6/20
89/89 [=======] - 45s 510ms/step -
loss: 0.6462 - accuracy: 0.7718 - val loss: 1383.3610 - val accuracy: 0.2616
Epoch 7/20
89/89 [=======] - 34s 387ms/step -
loss: 0.5867 - accuracy: 0.7928 - val_loss: 1626.8010 - val_accuracy:0.1771
Epoch 8/20
89/89 [=======] - 45s 504ms/step -
loss: 0.5461 - accuracy: 0.8058 - val_loss: 1733.9170 - val_accuracy:0.2014
Epoch 9/20
89/89 [=======] - 55s 617ms/step -
loss: 0.4965 - accuracy: 0.8283 - val loss: 2105.0442 - val accuracy: 0.2523
Epoch 10/20
```

```
89/89 [=======] - 55s 617ms/step -
loss: 0.5316 - accuracy: 0.8125 - val_loss: 1585.0485 - val_accuracy: 0.2766
Epoch 11/20
89/89 [======] - 52s 577ms/step -
loss: 0.5039 - accuracy: 0.8258 - val_loss: 1588.1725 - val_accuracy: 0.3032
Epoch 12/20
89/89 [=======] - 51s 571ms/step -
loss: 0.4196 - accuracy: 0.8546 - val loss: 2111.2288 - val accuracy: 0.2824
Epoch 13/20
loss: 0.4402 - accuracy: 0.8504 - val loss: 1728.3689 - val accuracy: 0.2824
Epoch 14/20
89/89 [=======] - 51s 568ms/step -
loss: 0.4035 - accuracy: 0.8560 - val_loss: 1953.9325 - val_accuracy:0.2477
Epoch 15/20
89/89 [=======] - 52s 578ms/step -
loss: 0.3994 - accuracy: 0.8606 - val_loss: 1739.5107 - val_accuracy: 0.2894
Epoch 16/20
89/89 [=======] - 51s 575ms/step -
loss: 0.3509 - accuracy: 0.8754 - val loss: 1912.0873 - val accuracy: 0.3252
Epoch 17/20
```

### MODEL BUILDING FOR VEGETABLE DISEASE PREDICTION

from keras.models import

Sequentialfrom keras.layers import

Dense

from keras.layers import Convolution2D

from keras.layers import MaxPooling2D

from keras.layers import Flatten

model=Sequential()

model.add(Convolution2D(32,(3,3),input\_shape =

(128,128,3), activation = 'relu'))

model.add(MaxPooling2D(pool\_size = (2,2)))

```
model.add(Flatten())
model.add(Dense(300, 'relu'))
model.add(Dense(150, 'relu'))
model.add(Dense(75, 'relu'))
model.add(Dense(9, 'softmax',
))
model.compile(optimizer='adam', loss = "categorical crossentropy",
metrics =['accuracy'])
model.summary()
model.fit(x train,epochs=20,steps per epoch=89,validation data
=x_{test}, validation_steps = 27)
model.save("veg.h5")
OUTPUT:
C:\Users\mumma\PycharmProjects\plant_disease_detection\venv\Sc
ripts\python.exe
C:\Users\mumma\PycharmProjects\plant_disease_detection\veg_m
odel.py
Found 11386 images belonging to 9 classes.
Found 3416 images belonging to 9 classes.
Model: "sequential"
Layer (type)
                          Output Shape
                                                           Param #
                           (None, 126, 126, 32)
conv2d (Conv2D)
                                                            896
```

max_pooling2d	max_	_pool	ling2d
---------------	------	-------	--------

(MaxPooling2D)

(None, 63, 63, 32)

0

flatten (Flatten)

(None, 127008)

0

dense (Dense)

(None, 300)

38102700

dense\_1 (Dense)

(None, 150)

45150

dense\_2 (Dense)

(None, 75)

11325

dense\_3 (Dense)

(None, 9)

684

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Total params: 38,160,755

Trainable params: 38,160,755

Non-trainable params: 0

# Epoch 1/20

89/89 [======] - 42s 459ms/step -

loss: 2.3594 - accuracy: 0.3153 - val\_loss: 194.4259 - val\_accuracy:

0.4931

```
Epoch 2/20
loss: 1.2180 - accuracy: 0.5829 - val loss: 532.0151 - val accuracy:
0.3472
Epoch 3/20
89/89 [=======] - 47s 527ms/step -
loss: 0.8772 - accuracy: 0.6822 - val loss: 699.0117 - val accuracy:
0.3669
Epoch 4/20
89/89 [=======] - 47s 531ms/step -
loss: 0.7295 - accuracy: 0.7468 - val loss: 1188.0234 - val accuracy: 0.2616
Epoch 5/20
loss: 0.6464 - accuracy: 0.7738 - val loss: 1339.5924 - val accuracy: 0.3241
Epoch 6/20
89/89 [=======] - 45s 510ms/step -
loss: 0.5844 - accuracy: 0.7949 - val_loss: 1505.8514 - val_accuracy:0.2477
Epoch 7/20
89/89 [=======] - 46s 514ms/step -
loss: 0.5777 - accuracy: 0.7960 - val_loss: 1878.0940 - val_accuracy: 0.2407
Epoch 8/20
89/89 [=======] - 45s 509ms/step -
loss: 0.5296 - accuracy: 0.8146 - val loss: 1235.0643 - val accuracy: 0.3264
Epoch 9/20
```

```
89/89 [=======] - 46s 518ms/step -
loss: 0.4724 - accuracy: 0.8332 - val_loss: 1428.7808 - val_accuracy:0.2662
Epoch 10/20
89/89 [======] - 41s 460ms/step -
loss: 0.4665 - accuracy: 0.8350 - val_loss: 1449.1967 - val_accuracy:0.2951
Epoch 11/20
89/89 [=======] - 44s 491ms/step -
loss: 0.4025 - accuracy: 0.8592 - val loss: 1373.3068 - val accuracy: 0.2581
Epoch 12/20
89/89 [=======] - 45s 508ms/step -
loss: 0.4034 - accuracy: 0.8574 - val loss: 1854.9781 - val accuracy: 0.2292
Epoch 13/20
89/89 [=======] - 49s 551ms/step -
loss: 0.3759 - accuracy: 0.8648 - val loss: 2173.3513 - val accuracy: 0.2188
Epoch 14/20
89/89 [=======] - 48s 539ms/step -
loss: 0.3696 - accuracy: 0.8687 - val_loss: 1862.3169 - val_accuracy:0.2431
Epoch 15/20
89/89 [======] - 49s 555ms/step -
loss: 0.3470 - accuracy: 0.8834 - val loss: 1886.7236 - val accuracy: 0.2245
Epoch 16/20
```