

<b>Date</b>	<b>07.11.2022</b>
<b>Team ID</b>	<b>PNT2022TMID22638</b>
<b>Project Name</b>	<b>FERTILIZERS RECOMMENDATION SYSTEM FOR DISEASE PREDICTION</b>

### **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:**

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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

=====

Total params: 38,160,755

Trainable params: 38,160,755

Non-trainable params: 0

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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

89/89 [=====] - 52s 582ms/step -  
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

89/89 [=====] - 50s 561ms/step

-

loss: 0.3818 - accuracy: 0.8606 - val\_loss: 1777.9532 -

val\_accuracy:0.3125

Epoch 18/20

89/89 [=====] - 50s 565ms/step

-

loss: 0.3416 - accuracy: 0.8810 - val\_loss: 2017.1232 -

val\_accuracy:0.2801

Epoch 19/20

89/89 [=====] - 51s 574ms/step

-

loss: 0.3515 - accuracy: 0.8743 - val\_loss: 1423.0455 -

val\_accuracy:0.3530

Epoch 20/20

89/89 [=====] - 50s 560ms/step

-

loss: 0.3514 - accuracy: 0.8761 - val\_loss: 1466.1351 -

val\_accuracy:0.3218