

```

from keras.preprocessing.image import ImageDataGenerator
train_datagen=ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_flip=True)
test_datagen=ImageDataGenerator(rescale=1)

x_train=train_datagen.flow_from_directory(r'/content/drive/MyDrive/Dataset/Dataset Plant Disease/fruit-dataset/fruit-dataset/test',target_size=(128,128),batch_size=2,class_mode='categorical')
x_test=test_datagen.flow_from_directory(r'/content/drive/MyDrive/Dataset/Dataset Plant Disease/fruit-dataset/fruit-dataset/train',target_size=(128,128),batch_size=2,class_mode='categorical')

Found 1686 images belonging to 6 classes.
Found 5384 images belonging to 6 classes.

```

1. 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

```

1. initializing the model

```
model=Sequential()
```

1. Add CNN layers

```
model.add(Convolution2D(32,
(3,3),input_shape=(128,128,3),activation='relu'))
```

```
model.add(MaxPooling2D(pool_size=(2,2)))
```

```
model.add(Flatten())
```

1. Add dense layer

```
model.add(Dense(units=40,kernel_initializer='uniform',activation='relu'))
```

```
model.add(Dense(units=20,kernel_initializer='random_uniform',activation='relu'))
```

```
model.add(Dense(units=6,kernel_initializer='random_uniform',activation='softmax'))
```

1. Train and save the model

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

```
model.fit(x_train,steps_per_epoch=89,epochs=20,validation_data=x_test,validation_steps=27)
```

Epoch 1/20
89/89 [=====] - 146s 2s/step - loss: 1.6616 - accuracy: 0.3764 - val_loss: 203.1930 - val_accuracy: 0.2963
Epoch 2/20
89/89 [=====] - 129s 1s/step - loss: 1.7158 - accuracy: 0.2697 - val_loss: 22.3784 - val_accuracy: 0.2778
Epoch 3/20
89/89 [=====] - 125s 1s/step - loss: 1.6271 - accuracy: 0.3258 - val_loss: 163.5451 - val_accuracy: 0.3333
Epoch 4/20
89/89 [=====] - 112s 1s/step - loss: 1.3890 - accuracy: 0.4888 - val_loss: 88.6855 - val_accuracy: 0.5926
Epoch 5/20
89/89 [=====] - 112s 1s/step - loss: 0.9276 - accuracy: 0.6236 - val_loss: 164.1111 - val_accuracy: 0.6667
Epoch 6/20
89/89 [=====] - 105s 1s/step - loss: 0.7846 - accuracy: 0.6798 - val_loss: 71.4850 - val_accuracy: 0.6481
Epoch 7/20
89/89 [=====] - 99s 1s/step - loss: 0.7925 - accuracy: 0.7135 - val_loss: 102.9553 - val_accuracy: 0.5926
Epoch 8/20
89/89 [=====] - 98s 1s/step - loss: 0.7527 - accuracy: 0.7135 - val_loss: 560.5753 - val_accuracy: 0.5000
Epoch 9/20
89/89 [=====] - 92s 1s/step - loss: 0.7694 - accuracy: 0.6966 - val_loss: 69.2323 - val_accuracy: 0.7963
Epoch 10/20
89/89 [=====] - 95s 1s/step - loss: 0.6303 - accuracy: 0.8090 - val_loss: 126.6944 - val_accuracy: 0.6296
Epoch 11/20
89/89 [=====] - 88s 978ms/step - loss: 0.6382 - accuracy: 0.7584 - val_loss: 65.5593 - val_accuracy: 0.7593
Epoch 12/20
89/89 [=====] - 87s 980ms/step - loss: 0.6182 - accuracy: 0.7865 - val_loss: 86.7426 - val_accuracy: 0.6667
Epoch 13/20
89/89 [=====] - 84s 938ms/step - loss: 0.5206 - accuracy: 0.8034 - val_loss: 43.7637 - val_accuracy: 0.8333
Epoch 14/20
89/89 [=====] - 86s 976ms/step - loss: 0.5636 - accuracy: 0.8202 - val_loss: 112.9079 - val_accuracy: 0.7037
Epoch 15/20
89/89 [=====] - 83s 937ms/step - loss: 0.5015 - accuracy: 0.8315 - val_loss: 81.1166 - val_accuracy: 0.7407
Epoch 16/20
89/89 [=====] - 84s 943ms/step - loss: 0.4755 - accuracy: 0.8315 - val_loss: 97.4727 - val_accuracy: 0.7593
Epoch 17/20
89/89 [=====] - 85s 965ms/step - loss: 0.4559

```
- accuracy: 0.8427 - val_loss: 88.8596 - val_accuracy: 0.7407
Epoch 18/20
89/89 [=====] - 82s 923ms/step - loss: 0.3686
- accuracy: 0.8596 - val_loss: 107.9981 - val_accuracy: 0.7222
Epoch 19/20
89/89 [=====] - 80s 901ms/step - loss: 0.4244
- accuracy: 0.8764 - val_loss: 34.6990 - val_accuracy: 0.8704
Epoch 20/20
89/89 [=====] - 80s 897ms/step - loss: 0.5965
- accuracy: 0.7809 - val_loss: 64.9681 - val_accuracy: 0.7222
```

```
<keras.callbacks.History at 0x7f2f0fc41d90>
```

```
model.save('fruit.h5')
```

```
model.summary()
```

```
Model: "sequential"
```

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, 40)	12040
dense_2 (Dense)	(None, 20)	820
dense_3 (Dense)	(None, 6)	126

```
=====  
Total params: 38,116,582  
Trainable params: 38,116,582  
Non-trainable params: 0  
=====
```