

```

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

from keras.preprocessing.image import ImageDataGenerator

train_datagen = ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal

test_datagen = ImageDataGenerator(rescale=1./255)

x_train = train_datagen.flow_from_directory("/content/drive/MyDrive/AI_IBM/Dataset/TRAIN_S

    Found 4119 images belonging to 5 classes. x_test =

test_datagen.flow_from_directory("/content/drive/MyDrive/AI_IBM/Dataset/TEST_SET"

    Found 929 images belonging to 5 classes.

x_train.class_indices

{'APPLES': 0, 'BANANA': 1, 'ORANGE': 2, 'PINEAPPLE': 3, 'WATERMELON': 4}

print(x_test.class_indices)

{'APPLES': 0, 'BANANA': 1, 'ORANGE': 2, 'PINEAPPLE': 3, 'WATERMELON': 4}

from collections import Counter as c c(x_train
.labels)

Counter({0: 995, 1: 1355, 2: 1019, 3: 275, 4: 475})

model = Sequential()

model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation="relu"))

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

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

model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten()) model.add(Dense(units=128,activation='relu'))

```

```
model.add(Dense(units=5,activation='softmax'))
```

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

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	9248
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 32)	0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 128)	802944
dense_1 (Dense)	(None, 5)	645
flatten_1 (Flatten)	(None, 5)	0

=====
Total params: 813,733
Trainable params: 813,733
Non-trainable params: 0
=====

```
model.add(Dense(units=300,kernel_initializer="random_uniform",activation="relu"))
```

```
model.add(Dense(units=200,kernel_initializer="random_uniform",activation="relu"))
```

```
model.add(Dense(units=5,kernel_initializer="random_uniform",activation="softmax"))  
len(x_train)
```

129

```
model.add(Dense(units=128,activation="relu",kernel_initializer="random_uniform"))
```

```
model.add(Dense(units=1,activation="sigmoid",kernel_initializer="random_uniform"))
```

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

```
model.fit_generator(x_train,steps_per_epoch=len(x_train), validation_data=x_test, validation_steps=10,  
                    /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: UserWarning: `Model.fit_generator`
```

```
                    """Entry point for launching an IPython kernel. Epoch
```

1/20

129/129 [=====] - 42s 323ms/step - loss: -579.1954 - accuracy: 0.9999
Epoch 2/20

129/129 [=====] - 35s 272ms/step - loss: -630.7393 - accuracy: 0.9999
Epoch 3/20

129/129 [=====] - 35s 273ms/step - loss: -683.9399 - accuracy: 0.9999
Epoch 4/20

129/129 [=====] - 35s 274ms/step - loss: -738.6011 - accuracy: 0.9999
Epoch 5/20

129/129 [=====] - 36s 275ms/step - loss: -795.0793 - accuracy: 0.9999
Epoch 6/20

```

129/129 [=====] - 37s 286ms/step - loss: -853.5035 -
accuracy Epoch 7/20
129/129 [=====] - 36s 276ms/step - loss: -913.4440 -
accuracy
Epoch 8/20
129/129 [=====] - 36s 275ms/step - loss: -974.8712 -
accuracy Epoch 9/20
129/129 [=====] - 35s 274ms/step - loss: -1037.6532 -
accuracy Epoch 10/20
129/129 [=====] - 36s 275ms/step - loss: -1101.9432 -
accuracy Epoch 11/20
129/129 [=====] - 35s 273ms/step - loss: -1167.7832 -
accuracy
Epoch 12/20
129/129 [=====] - 35s 273ms/step - loss: -1235.0177 -
accuracy Epoch 13/20
129/129 [=====] - 35s 274ms/step - loss: -1303.9956 -
accuracy Epoch 14/20
129/129 [=====] - 35s 274ms/step - loss: -1374.5148 -
accuracy
Epoch 15/20
129/129 [=====] - 36s 276ms/step - loss: -1446.9734 -
accuracy
Epoch 16/20
129/129 [=====] - 35s 274ms/step - loss: -1520.6868 -
accuracy Epoch 17/20
129/129 [=====] - 35s 273ms/step - loss: -1596.1498 -
accuracy Epoch 18/20
129/129 [=====] - 35s 271ms/step - loss: -1673.0337 -
accuracy
Epoch 19/20
129/129 [=====] - 35s 273ms/step - loss: -1751.5466 -
accuracy Epoch 20/20
129/129 [=====] - 35s 270ms/step - loss: -1831.8647 - accuracy
<keras.callbacks.History at 0x7f60240c4c10>

```

```
model.save("nutrition.h5")
```

```

from tensorflow.keras.models import load_model
from keras.preprocessing import image
model = load_model("nutrition.h5")

```

```
import numpy as np
```

```

from tensorflow.keras.utils import load_img
img = load_img(r'/content/drive/MyDrive/AI_IBM/Nutrition Analysis Using Image Classification')
x = img_to_array(img)

```

```

x = np.expand_dims(x, axis = 0)
predict_x = model.predict(x)
classes_x = np.argmax(predict_x, axis = -1)
classes_x

```

1/1 [=====] - 0s 424ms/step array([0])

```
index=['APPLES', 'BANANA', 'ORANGE', 'PINEAPPLE', 'WATERMELON']  
result=str(index[classes_x[0]]) result
```

'APPLES'