

AI Powered Nutrition Analyzer for Fitness Enthusiasts

Project development phase

Sprint-2

Date	12.11.2022
Team ID	PNT2022TMID37915
Project name	AI Powered Nutrition Analyzer for Fitness Enthusiasts

```
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_flip=True)
```

```
test_datagen = ImageDataGenerator(rescale=1./255)
```

```
x_train = train_datagen.flow_from_directory("/content/drive/MyDrive/AI_IBM/Dataset/TRAIN_SET",
```

```
    class_index_from_label={'APPLE': 0, 'BANANA': 1, 'ORANGE': 2, 'PINEAPPLE': 3, 'WATERMELON': 4})
```

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

```
    class_index_from_label={'APPLE': 0, 'BANANA': 1, 'ORANGE': 2, 'PINEAPPLE': 3, 'WATERMELON': 4})
```

```
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)
```

```

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=len(x_test),
                    /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: 0.9999
Epoch 7/20
129/129 [=====] - 36s 276ms/step - loss: -913.4440 - accuracy: 0.9999
Epoch 8/20
129/129 [=====] - 36s 275ms/step - loss: -974.8712 - accuracy: 0.9999
Epoch 9/20
129/129 [=====] - 35s 274ms/step - loss: -1037.6532 - accuracy: 0.9999
Epoch 10/20
129/129 [=====] - 36s 275ms/step - loss: -1101.9432 - accuracy: 0.9999
Epoch 11/20
129/129 [=====] - 35s 273ms/step - loss: -1167.7832 - accuracy: 0.9999
Epoch 12/20
129/129 [=====] - 35s 273ms/step - loss: -1235.0177 - accuracy: 0.9999
Epoch 13/20
129/129 [=====] - 35s 274ms/step - loss: -1303.9956 - accuracy: 0.9999
Epoch 14/20
129/129 [=====] - 35s 274ms/step - loss: -1374.5148 - accuracy: 0.9999
Epoch 15/20
129/129 [=====] - 36s 276ms/step - loss: -1446.9734 - accuracy: 0.9999
Epoch 16/20
129/129 [=====] - 35s 274ms/step - loss: -1520.6868 - accuracy: 0.9999
Epoch 17/20
129/129 [=====] - 35s 273ms/step - loss: -1596.1498 - accuracy: 0.9999
Epoch 18/20
129/129 [=====] - 35s 271ms/step - loss: -1673.0337 - accuracy: 0.9999
Epoch 19/20
129/129 [=====] - 35s 273ms/step - loss: -1751.5466 - accuracy: 0.9999
Epoch 20/20
129/129 [=====] - 35s 270ms/step - loss: -1831.8647 - accuracy: 0.9999
<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 = image.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'
```