

Team ID	PNT2022TMID38280
Project Name	AI - powered Nutrition Analyzer for Fitness Enthusiasts

### Model Building

```
[16]: # Importing Necessary Libraries import numpy as np #used for numerical analysis import
tensorflow #open source used for both ML and DL for computation from
tensorflow.keras.models import Sequential #it is a plain stack of Layers from tensorflow.keras
import layers #A Layer consists of a tensor-in tensor-out ,→computation function

#Dense Layer is the regular deeply connected neural network Layer from
tensorflow.keras.layers import Dense, Flatten
#Flatten-used for flattening the input or change the dimension from
tensorflow.keras.layers import Conv2D, MaxPooling2D, Dropout ,→#Convolutional layer

#MaxPooling2D-for downsampling the image from keras.preprocessing.image
import ImageDataGenerator
```

```
[17]: model=Sequential()
```

#### 3.0.1 Creating the model

```
[18]: # Initializing the CNN classifier =
Sequential()
# First convolution Layer and pooling classifier.add(Conv2D(32, (3, 3), input_shape=(64, 64,
3), activation='relu')) classifier.add(MaxPooling2D(pool_size=(2, 2))) # Second convolution
Layer and pooling classifier.add(Conv2D(32, (3, 3), activation='relu'))
# input_shape is going to be the pooled feature maps from the previous ,→convolution Layer

classifier.add(MaxPooling2D(pool_size=(2 2)))
# Flattening the Layers
classifier.add(Flatten())
```

```
[19]: classifier.add(Dense (units=128 activation='relu'))
classifier.add(Dense (units=5, activation='softmax )) # softmax for more than 2
```

```
[20]: classifier.summary()
```

Model: "sequential\_1"

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 (MaxPooling (None, 14, 14, 32) 2D)		0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 128)	802944
dense_1 (Dense)	(None, 5)	645
=====		
Total params: 813,733		
Trainable params: 813,733		
Non-trainable params: 0		

```
[21]: classifier.compile(optimizer='adam', loss='sparse_categorical_crossentropy',  
_,>metrics=['accuracy'])
```

### 3.1 Fitting the model

```
[22]: classifier.fit_generator( generator=x_train, steps_per_epoch = len(x_train),  
_,>epochs=20, validation_data=x_test, validation_steps = len(x_test)) # No of  
_,>images in test set
```

Epoch 1/20

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:1: UserWarning: `Model.fit\_generator` is deprecated and will be removed in a future version.

Please use `Model.fit`, which supports generators. """Entry point for launching an IPython kernel.

526/526 [=====] - 15s 13ms/step - loss: 0.1652 - accuracy: 0.9391 - val\_loss: 0.1162 - val\_accuracy: 0.9621

Epoch 2/20

526/526 [=====] - 6s 12ms/step - loss: 0.0592 - accuracy: 0.9817 - val\_loss: 0.0045 - val\_accuracy: 1.0000

Epoch 3/20

526/526 [=====] - 6s 12ms/step - loss: 4.5107e-04 - accuracy: 1.0000 - val\_loss: 0.0203 - val\_accuracy: 0.9896

Epoch 4/20

526/526 [=====] - 6s 12ms/step - loss: 1.8523e-04 - accuracy: 1.0000 - val\_loss: 0.0052 - val\_accuracy: 1.0000

Epoch 5/20

526/526 [=====] - 6s 12ms/step - loss: 8.9034e-05 - accuracy: 1.0000 - val\_loss: 0.0113 - val\_accuracy: 0.9905

Epoch 6/20

526/526 [=====] - 6s 12ms/step - loss:  
5.6083e-05 - accuracy: 1.0000 - val\_loss: 0.0066 - val\_accuracy: 1.0000  
Epoch 7/20  
526/526 [=====] - 8s 14ms/step - loss:  
3.1644e-05 - accuracy: 1.0000 - val\_loss: 0.0128 - val\_accuracy: 0.9905  
Epoch 8/20  
526/526 [=====] - 6s 12ms/step - loss:  
2.3077e-05 - accuracy: 1.0000 - val\_loss: 0.0188 - val\_accuracy: 0.9896  
Epoch 9/20  
526/526 [=====] - 6s 12ms/step - loss:  
2.8951e-05 - accuracy: 1.0000 - val\_loss: 0.0113 - val\_accuracy: 0.9915  
Epoch 10/20  
526/526 [=====] - 7s 13ms/step - loss:  
1.6114e-05 accuracy: 1.0000 - val\_loss: 0.0256 - val\_accuracy:  
0.9867  
Epoch 11/20  
526/526 [=====] - 7s 13ms/step - loss:  
1.4261e-05 accuracy: 1.0000 - val\_loss: 0.0124 - val\_accuracy:  
0.9905  
Epoch 12/20  
526/526 [=====] - 7s 13ms/step - loss:  
5.8428e-06 accuracy: 1.0000 - val\_loss: 0.0147 - val\_accuracy:  
0.9905  
Epoch 13/20  
526/526 [=====] - 7s 13ms/step - loss:  
4.0379e-06 accuracy: 1.0000 - val\_loss: 0.0121 - val\_accuracy:  
0.9915  
Epoch 14/20  
526/526 [=====] - 7s 13ms/step - loss:  
4.0424e-06 - accuracy: 1.0000 - val\_loss: 0.0118 - val\_accuracy: 0.9915  
Epoch 15/20  
526/526 [=====] - 7s 13ms/step - loss:  
2.0868e-06 accuracy: 1.0000 - val\_loss: 0.0140 - val\_accuracy:  
0.9905  
Epoch 16/20  
526/526 [=====] - 7s 13ms/step - loss:  
1.3716e-06 - accuracy: 1.0000 - val\_loss: 0.0019 - val\_accuracy: 1.0000  
Epoch 17/20  
526/526 [=====] - 8s 14ms/step - loss:  
1.5067e-06 - accuracy: 1.0000 - val\_loss: 0.0177 - val\_accuracy: 0.9896  
Epoch 18/20  
526/526 [=====] - 7s 13ms/step - loss:  
1.2072e-06 - accuracy: 1.0000 - val\_loss: 0.0248 - val\_accuracy: 0.9877  
Epoch 19/20  
526/526 [=====] - 7s 13ms/step - loss:  
7.0966e-07 - accuracy: 1.0000 - val\_loss: 0.0147 - val\_accuracy: 0.9905  
Epoch 20/20

```
526/526 [=====] - 7s 13ms/step - loss:
0.0510 accuracy: 0.9890 - val_loss: 5.1513e-04 - val_accuracy:
1.0000
```

[22]: <keras.callbacks.History at 0x7f4ed0fb4c50>

```
[23]: classifier.save('nutrition.h5')
```

Test the model

[24]: *### Predicting our results from tensorflow.keras.preprocessing*

```
import image from tensorflow.keras.models
```

```
import load_model #from
```

```
keras.preprocessing import image
```

```
model = load_model("nutrition.h5") #Loading the model
for testing
```

```
[25]: img = image.load_img('/content/TEST_SET/APPLES/151_100.jpg',
```

```
→target_size=(64,64)) #Loading of the image
```

```
#img =
```

```
image.load_img('/content/dataset/Testing/bears/k4
(88).
```

```
→jpeg',target_size=(64,64)) x =
```

```
image.img_to_array(img) #image to array x =
```

```
np.expand_dims(x,axis=0) #changing the shape
```

```
#pred = model.predict_classes(x) #predicting
the classes
```

```
#pred
```

```
pred = np.argmax(model.predict(x))
```

grayscale=False, \_

```
print(pred, model.predict(x))
```

```
1/1 [=====] - 0s 96ms/step
```

```
1/1 [=====] - 0s 18ms/step
```

```
0 [[1. 0. 0. 0. 0.]]
```

```
[26]: op ['APPLES', 'BANANA' 'ORANGE' 'PINEAPPLE' 'WATERMELON'] # Creating list
```

```
→ of output categories
```

```
result op[pred]
```

```
print(result)
```

APPLES

[29]: img =

```
image.load_img('/content/TEST_SET/WATERMELON/143_100.jpg', _
```

```
→, grayscale=False, target_size=(64,64)) #Loading of the image #img =
```

```

image.load_img('/content/dataset/Testing/bears/k4
(88) .
→jpeg',target_size=(64,64)) x =
image.img_to_array(img) #image to array x =
np.expand_dims(x,axis=0) #changing the shape
#pred = model.predict_classes(x) #predicting
the classes

```

```

#pred pred =
np.argmax(model.predict(x))
print(pred, model.predict(x))

```

```

1/1 [=====] - 0s 17ms/step
1/1 [=====] - 0s 17ms/step
4 [[0. 0. 0. 0. 1.]]

```

```

[30]: op    ['APPLES', 'BANANA' 'ORANGE' 'PINEAPPLE' 'WATERMELON ] # Creating list
→ of output categories
result    op[pred]
print(result)

```

WATERMELON