

## Sprint-2

### Model Building (Training, Saving, Testing the model)

Date	11 November 2022
Team ID	PNT2022TMID09928
Project Name	AI-powered Nutrition Analyzer for Fitness Enthusiasts
Maximum Marks	

#### **Dataset:**

- In our dataset we have collected images of the five varieties of fruits.
  - Apple
  - Orange
  - Pineapple
  - Watermelon
  - Banana

Drive link : [https://drive.google.com/file/d/1jzDjV7jYclzllieagaJdubMJ3YeLsry1/view?usp=share\\_link](https://drive.google.com/file/d/1jzDjV7jYclzllieagaJdubMJ3YeLsry1/view?usp=share_link)

#### **Image Pre-processing:**

- Import The ImageDataGenerator Library
- Configure ImageDataGenerator Class
- Apply Image DataGenerator Functionality To Trainset And Testset

#### **Model Building:**

- Importing The Model Building Libraries
- Initializing The Model
- Adding CNN Layers
- Adding Dense Layers
- Configure The Learning Process
- Train the model
- Save the model
- Test the model

## ▼ Data Collection

Download the dataset [here](#)

# Unzipping the dataset

```
!unzip '/content/Dataset.zip'
```

```
inflating: Dataset/TRAIN_SET/WATERMELON/r_288_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_289_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_28_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_290_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_291_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_292_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_293_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_294_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_295_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_296_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_297_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_298_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_299_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_29_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_2_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_300_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_301_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_302_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_303_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_304_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_305_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_306_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_307_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_308_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_309_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_30_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_310_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_311_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_312_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_313_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_314_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_315_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_31_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_32_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_33_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_34_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_35_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_36_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_37_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_38_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_39_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_3_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_40_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_41_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_42_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_43_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_44_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_45_100.jpg
```

```
inflating: Dataset/TRAIN_SET/WATERMELON/r_46_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_4_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_50_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_57_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_5_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_6_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_7_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_81_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_8_100.jpg
inflating: Dataset/TRAIN_SET/WATERMELON/r_9_100.jpg
```

## ▼ Image Preprocessing

```
#Importing The ImageDataGenerator Library
from keras.preprocessing.image import ImageDataGenerator
```

## ▼ Image Data Augmentation

```
#Configure ImageDataGenerator Class
train_datagen = ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=True)
test_datagen=ImageDataGenerator(rescale=1./255)
```

## ▼ Applying Image DataGenerator Functionality To Trainset And Testset

```
#Applying Image DataGenerator Functionality To Trainset And Testset
x_train = train_datagen.flow_from_directory(
    r'/content/Dataset/TRAIN_SET',
    target_size=(64, 64), batch_size=5, color_mode='rgb', class_mode='sparse')
#Applying Image DataGenerator Functionality To Testset
x_test = test_datagen.flow_from_directory(
    r'/content/Dataset/TEST_SET',
    target_size=(64, 64), batch_size=5, color_mode='rgb', class_mode='sparse')

Found 4118 images belonging to 5 classes.
Found 929 images belonging to 5 classes.

#checking the number of classes
print(x_train.class_indices)

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

#checking the number of classes
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: 1354, 2: 1019, 3: 275, 4: 475})
```

## ▼ Model Building

### 1. Importing The Model Building Libraries

```
import numpy as np  
import tensorflow as tf  
from tensorflow.keras.models import Sequential  
from tensorflow.keras import layers  
from tensorflow.keras.layers import Dense, Flatten  
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dropout
```

### 2. Initializing The Model

```
model = Sequential()
```

### 3. Adding CNN Layers

```
# 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())
```

### 4. Adding Dense Layers

```
classifier.add(Dense(units=128, activation='relu'))
```

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

```
#summary of our model
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 (MaxPooling2D)	(None, 14, 14, 32)	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  
=====

## 5. Configure The Learning Process

```
# Compiling the CNN
# categorical_crossentropy for more than 2
classifier.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['acc
```

## 6. Train The Model

```
#Fitting the model
classifier.fit_generator(generator=x_train,steps_per_epoch = len(x_train),epochs=20, valid
```

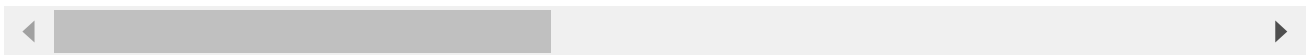
```
Epoch 1/20
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: UserWarning: `Model.

824/824 [=====] - 21s 16ms/step - loss: 0.6172 - accuracy:
Epoch 2/20
824/824 [=====] - 13s 15ms/step - loss: 0.4115 - accuracy:
Epoch 3/20
824/824 [=====] - 13s 16ms/step - loss: 0.3766 - accuracy:
Epoch 4/20
824/824 [=====] - 13s 16ms/step - loss: 0.3484 - accuracy:
Epoch 5/20
```

```

824/824 [=====] - 13s 16ms/step - loss: 0.3243 - accuracy:
Epoch 6/20
824/824 [=====] - 13s 16ms/step - loss: 0.3240 - accuracy:
Epoch 7/20
824/824 [=====] - 13s 16ms/step - loss: 0.2887 - accuracy:
Epoch 8/20
824/824 [=====] - 13s 16ms/step - loss: 0.2728 - accuracy:
Epoch 9/20
824/824 [=====] - 13s 16ms/step - loss: 0.2717 - accuracy:
Epoch 10/20
824/824 [=====] - 14s 17ms/step - loss: 0.2365 - accuracy:
Epoch 11/20
824/824 [=====] - 13s 15ms/step - loss: 0.2301 - accuracy:
Epoch 12/20
824/824 [=====] - 13s 15ms/step - loss: 0.2083 - accuracy:
Epoch 13/20
824/824 [=====] - 13s 15ms/step - loss: 0.2049 - accuracy:
Epoch 14/20
824/824 [=====] - 12s 15ms/step - loss: 0.1930 - accuracy:
Epoch 15/20
824/824 [=====] - 13s 15ms/step - loss: 0.1807 - accuracy:
Epoch 16/20
824/824 [=====] - 13s 15ms/step - loss: 0.1712 - accuracy:
Epoch 17/20
824/824 [=====] - 13s 15ms/step - loss: 0.1599 - accuracy:
Epoch 18/20
824/824 [=====] - 13s 15ms/step - loss: 0.1619 - accuracy:
Epoch 19/20
824/824 [=====] - 13s 15ms/step - loss: 0.1505 - accuracy:
Epoch 20/20
824/824 [=====] - 12s 15ms/step - loss: 0.1211 - accuracy:
<keras.callbacks.History at 0x7fd655833d90>

```



## 7. Saving The Model

```
classifier.save('nutrition.h5')
```

## 8. Testing The Model

```

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

from tensorflow.keras.utils import img_to_array
#loading of the image
img = load_img(r'/content/Sample_Images/Test_Image1.jpg', grayscale=False, target_size= (64,
#image to array
x = img_to_array(img)
#changing the shape
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 18ms/step
array([0])
```

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

```
'APPLES'
```