Sprint-2

Model Building(Training,Saving,Testing the model)

Date	01 November 2022
Team ID	PNT2022TMID44824
Project Name	AI-powered Nutrition Analyzer for Fitness Enthusiasts
Maximum Marks	

Dataset:

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

Drive link :
https://drive.google.com/file/d/1jzDjV7jYcIzlIieagaJdubMJ3YeLsry1/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
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  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
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  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,horizonta
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 (MaxPooling  (None, 14, 14, 32)        0
2D)

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
```