# **Nutrition Image Analysis using CNN**

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Project Name	Al-Powered Nutrition Analyzer for Fitness Enthusiasts

#### Nutrition Image Analysis using CNN

```
In[3]:
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
In [4]:
ls
drive/
        sample data/
In [5]:
cd//content/drive/MyDrive/Colab Notebooks/Dataset
/content/drive/.shortcut-targets-by-id/1LL5lvl6AsdVwW9LWVu GXEUCoV7jYm-c/Data
In [6]:
IBM review.pptx photo-1589820296156-2454bb8a6ad1.jpg TRAIN SET/
nutrition.h5
                 TEST SET/
```

# **Importing Neccessary Libraries**

In [7]:

```
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
#Faltten-used fot 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
Image Data Agumentation
In [8]:
#setting parameter for Image Data agumentation to the training data
train datagen =
ImageDataGenerator(rescale=1./255, shear range=0.2, zoom range=0.2, horizontal
flip=True)
 #Image Data agumentation to the testing data
test datagen=ImageDataGenerator(rescale=1./255)
Loading our data and performing data agumentation
In [9]:
 #performing data agumentation to train data
x train = train datagen.flow from directory(
    r'/content/drive/MyDrive/Colab Notebooks/Dataset/TRAIN SET',
    target size=(64, 64),batch size=5,color mode='rgb',class mode='sparse')
 #performing data agumentation to test data
x test = test datagen.flow from directory(
    r'/content/drive/MyDrive/Colab Notebooks/Dataset/TEST SET',
    target size=(64, 64),batch size=5,color mode='rgb',class mode='sparse')
Found 4138 images belonging to 5 classes.
Found 929 images belonging to 3 classes.
In [10]:
print(x train.class indices) #checking the number of classes
{'APPLES': 0, 'BANANA': 1, 'ORANGE': 2, 'PINEAPPLE': 3, 'WATERMELON': 4}
In [11]:
print(x test.class indices) #checking the number of classes
{'APPLES': 0, 'BANANA': 1, 'ORANGE': 2}
In [12]:
from collections import Counter as c
c(x train .labels)
Out[12]:
Counter({0: 995, 1: 1374, 2: 1019, 3: 275, 4: 475})
Creating the model
In [13]:
```

# 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())

# Adding a fully connected layer
classifier.add(Dense(units=128, activation='relu'))
classifier.add(Dense(units=5, activation='softmax')) # softmax for more than
2
```

## In [14]:

classifier.summary() #summary of our model

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 32)	896
<pre>max_pooling2d (MaxPooling2D )</pre>	(None, 31, 31, 32)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	9248
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 14, 14, 32)	0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 128)	802944
dense_1 (Dense)	(None, 5)	645

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

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# Compiling the model

### In [15]:

```
# Compiling the CNN
```

<sup>#</sup> categorical\_crossentropy for more than 2

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

## Fitting the model

```
In [16]:
```

```
classifier.fit generator(
       generator=x train, steps per epoch = len(x train),
       epochs=10, validation data=x test, validation steps = len(x test)) #
No of images in test set
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:3: UserWarning:
`Model.fit generator` is deprecated and will be removed in a future version.
Please use `Model.fit`, which supports generators.
 This is separate from the ipykernel package so we can avoid doing imports u
ntil
Epoch 1/10
828/828 [============ ] - 1580s 2s/step - loss: 0.6022 - acc
uracy: 0.7608 - val loss: 0.6050 - val accuracy: 0.7621
Epoch 2/10
uracy: 0.8415 - val loss: 0.4744 - val accuracy: 0.8149
Epoch 3/10
uracy: 0.8579 - val loss: 0.4508 - val accuracy: 0.8127
Epoch 4/10
828/828 [============ ] - 50s 61ms/step - loss: 0.3606 - acc
uracy: 0.8594 - val loss: 0.4128 - val accuracy: 0.8471
Epoch 5/10
828/828 [============= ] - 51s 61ms/step - loss: 0.3412 - acc
uracy: 0.8743 - val loss: 0.4203 - val accuracy: 0.8321
Epoch 6/10
uracy: 0.8729 - val loss: 0.4781 - val accuracy: 0.8084
828/828 [============= ] - 51s 62ms/step - loss: 0.3006 - acc
uracy: 0.8859 - val_loss: 0.4085 - val_accuracy: 0.8461
Epoch 8/10
828/828 [============= ] - 52s 63ms/step - loss: 0.2810 - acc
uracy: 0.8862 - val loss: 0.6500 - val accuracy: 0.8073
828/828 [============ ] - 50s 60ms/step - loss: 0.2838 - acc
uracy: 0.8925 - val loss: 0.4216 - val accuracy: 0.8332
Epoch 10/10
828/828 [============= ] - 52s 63ms/step - loss: 0.2580 - acc
uracy: 0.9016 - val loss: 0.3874 - val accuracy: 0.8439
Out[16]:
```

## Saving our model

```
In [17]:
```

```
# Save the model
classifier.save('nutrition.h5')
```

# **Nutrition Image Analysis using CNN**

## **Predicting our results**

```
In [18]:
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
import numpy as np
In [19]:
img = image.load img("/content/drive/MyDrive/Colab
Notebooks/Dataset/TRAIN SET/APPLES/n07740461 10067.jpg",target size=
 (64,64)) #loading of the image
img
Out[19]:
In [20]:
```

```
x=image.img to array(img) #conversion image into array
In [21]:
```

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#### Out[21]:

```
array([[[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
```

```
. . . ,
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        . . . ,
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]],
       [[255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.],
        [255., 255., 255.]]], dtype=float32)
In [22]:
x.ndim
Out[22]:
In [23]:
x=np.expand_dims(x,axis=0) #expand the dimension
In [24]:
x.ndim
Out[24]:
4
In [25]:
pred = classifier.predict(x)
1/1 [======] - 0s 125ms/step
In [26]:
pred
Out[26]:
array([[1., 0., 0., 0., 0.]], dtype=float32)
```

```
In [27]:
    labels=['APPLES', 'BANANA', 'ORANGE', 'PINEAPPLE', 'WATERMELON']
    labels[np.argmax(pred)]
Out[27]:
    'APPLES'
```