

# UNZIP THE FILE

```
[ ] from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

/content/drive/MyDrive/Colab (ctrl + click)
lunzip '/content/drive/MyDrive/Colab Notebooks/TRAIN_SET.zip'

Archives: /content/drive/MyDrive/Colab Notebooks/TRAIN_SET.zip
replace TRAIN_SET/APPLES/0_100.jpg? [y]es, [n]o, [A]ll, [N]one, [r]ename: n
replace TRAIN_SET/APPLES/1_100.jpg? [y]es, [n]o, [A]ll, [N]one, [r]ename: a
error: invalid response [a]
replace TRAIN_SET/APPLES/1_100.jpg? [y]es, [n]o, [A]ll, [N]one, [r]ename: y
inflating: TRAIN_SET/APPLES/1_100.jpg
replace TRAIN_SET/APPLES/10_100.jpg? [y]es, [n]o, [A]ll, [N]one, [r]ename: A
inflating: TRAIN_SET/APPLES/10_100.jpg
inflating: TRAIN_SET/APPLES/100_100.jpg
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inflating: TRAIN_SET/APPLES/102_100.jpg
inflating: TRAIN_SET/APPLES/103_100.jpg
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inflating: TRAIN_SET/APPLES/106_100.jpg
inflating: TRAIN_SET/APPLES/107_100.jpg
inflating: TRAIN_SET/APPLES/108_100.jpg
inflating: TRAIN_SET/APPLES/109_100.jpg
inflating: TRAIN_SET/APPLES/11_100.jpg
inflating: TRAIN_SET/APPLES/110_100.jpg
inflating: TRAIN_SET/APPLES/111_100.jpg
inflating: TRAIN_SET/APPLES/112_100.jpg
inflating: TRAIN_SET/APPLES/113_100.jpg
inflating: TRAIN_SET/APPLES/114_100.jpg
inflating: TRAIN_SET/APPLES/115_100.jpg
inflating: TRAIN_SET/APPLES/116_100.jpg
```

```
Code * Text
inflating: TRAIN_SET/ORANGE/r_312_100.jpg
inflating: TRAIN_SET/ORANGE/r_313_100.jpg
inflating: TRAIN_SET/ORANGE/r_314_100.jpg
inflating: TRAIN_SET/ORANGE/r_315_100.jpg
inflating: TRAIN_SET/ORANGE/r_316_100.jpg
inflating: TRAIN_SET/ORANGE/r_317_100.jpg
inflating: TRAIN_SET/ORANGE/r_318_100.jpg
inflating: TRAIN_SET/ORANGE/r_319_100.jpg
inflating: TRAIN_SET/ORANGE/r_320_100.jpg
inflating: TRAIN_SET/PINEAPPLE/0_100.jpg
inflating: TRAIN_SET/PINEAPPLE/1_100.jpg
inflating: TRAIN_SET/PINEAPPLE/10_100.jpg
inflating: TRAIN_SET/PINEAPPLE/100_100.jpg
inflating: TRAIN_SET/PINEAPPLE/101_100.jpg
inflating: TRAIN_SET/PINEAPPLE/102_100.jpg
inflating: TRAIN_SET/PINEAPPLE/103_100.jpg
inflating: TRAIN_SET/PINEAPPLE/104_100.jpg
inflating: TRAIN_SET/PINEAPPLE/105_100.jpg
inflating: TRAIN_SET/PINEAPPLE/106_100.jpg
inflating: TRAIN_SET/PINEAPPLE/107_100.jpg
inflating: TRAIN_SET/PINEAPPLE/108_100.jpg
inflating: TRAIN_SET/PINEAPPLE/109_100.jpg
inflating: TRAIN_SET/PINEAPPLE/11_100.jpg
inflating: TRAIN_SET/PINEAPPLE/110_100.jpg
inflating: TRAIN_SET/PINEAPPLE/111_100.jpg
inflating: TRAIN_SET/PINEAPPLE/112_100.jpg
inflating: TRAIN_SET/PINEAPPLE/113_100.jpg
inflating: TRAIN_SET/PINEAPPLE/114_100.jpg
inflating: TRAIN_SET/PINEAPPLE/115_100.jpg
inflating: TRAIN_SET/PINEAPPLE/116_100.jpg
inflating: TRAIN_SET/PINEAPPLE/117_100.jpg

- Seperate as Test Set and Train Set
```

# SEPARATE THE TEST AND TRAIN SET

## Seperate as Test Set and Train Set

```
[ ] from keras.preprocessing.image import ImageDataGenerator
import tensorflow as tf

[ ] image_generator = tf.keras.preprocessing.image.ImageDataGenerator(validation_split=0.2)
train_data_gen = image_generator.flow_from_directory(directory='TRAIN_SET', subset='training')
val_data_gen = image_generator.flow_from_directory(directory='TRAIN_SET', subset='validation')

Found 2102 images belonging to 5 classes.
Found 524 images belonging to 5 classes.

[ ] 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(
    r'/content/TRAIN_SET',
    target_size=(64,64), batch_size=5, color_mode='rgb', class_mode='sparse')

x_test=test_datagen.flow_from_directory(
    r'/content/TRAIN_SET',
    target_size=(64,64), batch_size=5, color_mode='rgb', class_mode='sparse')

Found 2626 images belonging to 5 classes.
Found 2626 images belonging to 5 classes.
```

```
[ ] print(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: 606, 1: 445, 2: 479, 3: 621, 4: 475})
```

## Importing Neccesarry Libraries

```
[ ] import numpy as np
import tensorflow
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
from keras.preprocessing.image import ImageDataGenerator
```

# INITIALIZING & CREATING THE MODEL

```

- Initializing The Model

[ ] model=Sequential()

- Creating the model

[ ] classifier=Sequential()
  classifier.add(Conv2D(32,(3,3),input_shape=(64,64,3),activation='relu'))
  classifier.add(MaxPooling2D(pool_size=(2,2)))
  classifier.add(Conv2D(32,(3,3),activation='relu'))
  classifier.add(MaxPooling2D(pool_size=(2,2)))
  classifier.add(Flatten())

[ ] classifier.add(Dense(units=128, activation='relu'))
  classifier.add(Dense(units=5, activation='softmax'))

[ ] 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

```

```

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

[ ] 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

- Compiling the model

```

# COMPILING & FITTING THE MODEL

```
Compiling the model

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

Fitting the model

1 classifier.fit_generator(
    generator=x_train,steps_per_epoch=len(x_train),
    epochs=20,validation_data=x_test,validation_steps=len(x_test))

Epoch 1/20
/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
This is separate from the ipykernel package so we can avoid doing imports until
526/526 [=====] - 31s 57ms/step - loss: 1.6057 - accuracy: 0.2327 - val_loss: 1.6027 - val_accuracy: 0.2365
Epoch 2/20
526/526 [=====] - 30s 57ms/step - loss: 1.6019 - accuracy: 0.2281 - val_loss: 1.6007 - val_accuracy: 0.2365
Epoch 3/20
526/526 [=====] - 32s 61ms/step - loss: 1.6007 - accuracy: 0.2365 - val_loss: 1.6001 - val_accuracy: 0.2365
Epoch 4/20
526/526 [=====] - 29s 55ms/step - loss: 1.6004 - accuracy: 0.2239 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 5/20
526/526 [=====] - 28s 53ms/step - loss: 1.6003 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 6/20
526/526 [=====] - 28s 54ms/step - loss: 1.6003 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 7/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 8/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 9/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 10/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 11/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 12/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 13/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 14/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 15/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 16/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 17/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 18/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 19/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
Epoch 20/20
526/526 [=====] - 30s 58ms/step - loss: 1.6002 - accuracy: 0.2365 - val_loss: 1.5999 - val_accuracy: 0.2365
```

# SAVING THE MODEL & PREDICTING THE RESULTS

```
saving model

[ ] classifier.save('nutrition.h5')

Predicting results

1 import tensorflow
  from tensorflow.keras.models import load_model
  from keras.preprocessing import image
  from tensorflow.keras.utils import load_img,img_to_array
  model =load_model("nutrition.h5")

2 img= tensorflow.keras.utils.load_img(r"/content/TRAIN SET/APPLES/10_100.jpg",
    grayscale=False,target size= (64,64))
  x=tensorflow.keras.utils.img_to_array(img)
  x=np.expand_dims(x,axis = 0)
  os.pred= model.predict_classes(x)
  os.pred

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

