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

Mounted at /content/drive

import os
import random
import matplotlib.pyplot as plt
from PIL import Image

train_path = '/content/drive/MyDrive/AI &
ML/Week_5/FruitinAmazon/train'
test_path = '/content/drive/MyDrive/AI & ML/Week_5/FruitinAmazon/test'
```

## 3. Task - 1:

Repeat all the task from worksheet - 5 but, try to improve the model from last week with same dataset.

- Use Data Augmentation to increase the number of training image.
- Use deeper model with BN and DropOut layer as presented above.
- Understand the Model Summary and Training Behavior.

```
import tensorflow as tf
from tensorflow.keras import layers, models, regularizers
import matplotlib.pyplot as plt
train_dir = '/content/drive/MyDrive/AI &
ML/Week 5/FruitinAmazon/train'
img_height, img_width = 128, 128
batch size = 32
train ds = tf.keras.preprocessing.image dataset from directory(
    train dir,
    image size=(img height, img width),
    batch size=batch size,
    label mode='int',
    validation split=0.2,
    subset='training',
    seed=123
)
```

```
val ds = tf.keras.preprocessing.image dataset from directory(
    train dir,
    image_size=(img_height, img_width),
    batch size=batch size,
    label mode='int',
    validation_split=0.2,
    subset='validation',
    seed=123
)
class names = train ds.class names
print("Class names:", class names)
data augmentation = tf.keras.Sequential([
    layers.RandomFlip("horizontal"),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1),
    layers.RandomContrast(0.1),
])
def create improved model(input shape=(128, 128, 3), num classes=6):
    model = models.Sequential([
        layers.Input(shape=input shape),
        data augmentation,
        layers. Rescaling (1./255),
        layers.Conv2D(32, (3, 3), padding='same'),
        layers.BatchNormalization(),
        layers.Activation('relu'),
        layers.MaxPooling2D((2, 2)),
        layers.Dropout(0.25),
        layers.Conv2D(64, (3, 3), padding='same'),
        layers.BatchNormalization(),
        layers.Activation('relu'),
        layers.MaxPooling2D((2, 2)),
        layers.Dropout(0.25),
        layers.Conv2D(128, (3, 3), padding='same'),
        layers.BatchNormalization(),
        layers.Activation('relu'),
        layers.MaxPooling2D((2, 2)),
        layers.Dropout(0.25),
        layers.Conv2D(256, (3, 3), padding='same'),
        layers.BatchNormalization(),
        layers.Activation('relu'),
        layers.GlobalAveragePooling2D(),
        layers.Dropout(0.5),
```

```
layers.Dense(512, activation='relu',
kernel regularizer=regularizers.l2(0.01)),
        layers.BatchNormalization(),
        layers.Dropout(0.5),
        layers.Dense(num classes, activation='softmax')
    ])
    return model
improved model = create improved model(
    input shape=(img height, img width, 3),
    num classes=len(class names)
)
improved model.compile(
    optimizer=tf.keras.optimizers.Adam(learning rate=0.0001),
    loss='sparse_categorical_crossentropy',
    metrics=['accuracy']
)
improved model.summary()
history = improved model.fit(
    train ds,
    validation data=val ds,
    epochs=100,
    callbacks=[
        tf.keras.callbacks.ModelCheckpoint('best model.h5',
save_best_only=True),
        tf.keras.callbacks.EarlyStopping(patience=15)
    ]
)
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Val Accuracy')
plt.legend()
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Train Loss')
plt.plot(history.history['val loss'], label='Val Loss')
plt.legend()
plt.show()
Found 90 files belonging to 6 classes.
Using 72 files for training.
Found 90 files belonging to 6 classes.
Using 18 files for validation.
```

```
Class names: ['acai', 'cupuacu', 'graviola', 'guarana', 'pupunha',
'tucuma']
Model: "sequential 4"
Layer (type)
                                      Output Shape
Param #
sequential_3 (Sequential)
                                      (None, 128, 128, 3)
 rescaling 1 (Rescaling)
                                      (None, 128, 128, 3)
 conv2d 4 (Conv2D)
                                      (None, 128, 128, 32)
896
| batch_normalization 5
                                      (None, 128, 128, 32)
128
 (BatchNormalization)
 activation_4 (Activation)
                                      (None, 128, 128, 32)
0
 max pooling2d 3 (MaxPooling2D)
                                      (None, 64, 64, 32)
0 |
 dropout_5 (Dropout)
                                      (None, 64, 64, 32)
conv2d 5 (Conv2D)
                                      (None, 64, 64, 64)
18,496
 batch normalization 6
                                      (None, 64, 64, 64)
  (BatchNormalization)
```

<pre>  activation_5 (Activation) 0  </pre>	(None, 64, 64, 64)
max_pooling2d_4 (MaxPooling2D) 0	(None, 32, 32, 64)
dropout_6 (Dropout)	(None, 32, 32, 64)
conv2d_6 (Conv2D) 73,856	(None, 32, 32, 128)
batch_normalization_7 512     (BatchNormalization)	(None, 32, 32, 128)
activation_6 (Activation) 0	(None, 32, 32, 128)
max_pooling2d_5 (MaxPooling2D) 0	(None, 16, 16, 128)
dropout_7 (Dropout)	(None, 16, 16, 128)
conv2d_7 (Conv2D) 295,168	(None, 16, 16, 256)
batch_normalization_8 1,024   (BatchNormalization)	(None, 16, 16, 256)
activation_7 (Activation) 0	(None, 16, 16, 256)

```
global average pooling2d 1
                                       (None, 256)
  (GlobalAveragePooling2D)
 dropout 8 (Dropout)
                                        (None, 256)
0
 dense 2 (Dense)
                                        (None, 512)
131,584
 batch normalization 9
                                       (None, 512)
2,048
  (BatchNormalization)
 dropout_9 (Dropout)
                                       (None, 512)
 dense 3 (Dense)
                                       (None, 6)
3,078
Total params: 527,046 (2.01 MB)
Trainable params: 525,062 (2.00 MB)
Non-trainable params: 1,984 (7.75 KB)
Epoch 1/100
3/3 —
                    —— 0s 2s/step - accuracy: 0.1279 - loss: 6.2653
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
                    ——— 14s 2s/step - accuracy: 0.1411 - loss:
6.1991 - val accuracy: 0.1111 - val loss: 5.2077
Epoch 2/100
                    --- 0s 2s/step - accuracy: 0.1892 - loss: 5.9016
3/3 —
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save model(model)`. This file format
```

```
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my model.keras')` or
`keras.saving.save_model(model, 'my_model.keras')`.
              ______ 11s 2s/step - accuracy: 0.1940 - loss:
5.8996 - val accuracy: 0.1111 - val loss: 5.2036
Epoch 3/100
                 ———— 0s 1s/step - accuracy: 0.1852 - loss: 6.1824
3/3 —
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
                    ---- 6s 2s/step - accuracy: 0.1840 - loss: 6.1791
- val accuracy: 0.1111 - val loss: 5.2006
Epoch 4/100
                 ———— Os 1s/step - accuracy: 0.2043 - loss: 6.0135
3/3 —
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
                _____ 10s 2s/step - accuracy: 0.2088 - loss:
6.0123 - val accuracy: 0.1111 - val loss: 5.1978
Epoch 5/100
                 Os 1s/step - accuracy: 0.2818 - loss: 5.5803
3/3 ——
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
           _____ 11s 2s/step - accuracy: 0.2704 - loss:
5.5922 - val accuracy: 0.1111 - val loss: 5.1967
Epoch 6/100
               Os 2s/step - accuracy: 0.2396 - loss: 6.0203
3/3 —
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
             11s 2s/step - accuracy: 0.2422 - loss:
6.0200 - val accuracy: 0.1111 - val loss: 5.1961
```

```
Epoch 7/100
                ———— Os 1s/step - accuracy: 0.3409 - loss: 5.5447
3/3 —
WARNING:absl:You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save_model(model)`. This file format
is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
                9s 2s/step - accuracy: 0.3355 - loss: 5.5487
- val accuracy: 0.1111 - val loss: 5.1956
Epoch 8/100
                  ———— 10s 2s/step - accuracy: 0.2886 - loss: 5.3100
- val accuracy: 0.1111 - val loss: 5.1957
Epoch 9/100
                   —— 11s 2s/step - accuracy: 0.2943 - loss: 5.6821
3/3 —
- val accuracy: 0.1111 - val loss: 5.1960
Epoch 10/100
3/3 —
                 6s 2s/step - accuracy: 0.3720 - loss: 5.2126
- val accuracy: 0.1111 - val loss: 5.1965
Epoch 11/100 10s 2s/step - accuracy: 0.3099 - loss: 5.3117
- val accuracy: 0.1111 - val loss: 5.1972
Epoch 12/100
               _____ 10s 2s/step - accuracy: 0.2726 - loss: 5.6470
3/3 ———
- val accuracy: 0.1111 - val loss: 5.1981
Epoch 13/100
                  ----- 12s 2s/step - accuracy: 0.2483 - loss: 5.2958
3/3 ——
- val accuracy: 0.1111 - val loss: 5.1989
Epoch 14/100
                    — 9s 2s/step - accuracy: 0.3633 - loss: 5.2877
- val accuracy: 0.1111 - val_loss: 5.1997
Epoch 15/100
3/3 –
                     — 10s 2s/step - accuracy: 0.3446 - loss: 5.1824
- val accuracy: 0.1111 - val loss: 5.2008
Epoch 16/100
                8s 2s/step - accuracy: 0.4132 - loss: 5.0935
3/3 —
- val accuracy: 0.1111 - val loss: 5.2029
Epoch 17/100

8s 2s/step - accuracy: 0.3780 - loss: 5.2387
- val accuracy: 0.1111 - val_loss: 5.2058
Epoch 18/100
               _____ 10s 2s/step - accuracy: 0.3976 - loss: 5.1454
3/3 ———
- val accuracy: 0.1111 - val loss: 5.2093
Epoch 19/100
                11s 2s/step - accuracy: 0.4427 - loss: 4.8793
- val accuracy: 0.1111 - val loss: 5.2138
Epoch 20/100
                 ------ 11s 2s/step - accuracy: 0.3611 - loss: 5.5412
3/3 ——
- val accuracy: 0.1111 - val loss: 5.2200
```



