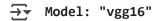
```
Start coding or generate with AI.
import tensorflow as tf
print("GPU Available:", tf.config.list physical devices('GPU'))
    GPU Available: []
!nvidia-smi
     /bin/bash: line 1: nvidia-smi: command not found
with tf.device('/GPU:0'):
  print("GPU Available:", tf.config.list physical devices('GPU'))
    GPU Available: []
import tensorflow as tf
from tensorflow import keras
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
# import zipfile
# zip_ref = zipfile.ZipFile('/content/drive/MyDrive/Level 6/Artificial_Intelligence/Week5/FruitinAmazon.zip', 'r')
```

```
# zip ref.extractall('/content')
# zip ref.close()
# data.data.districts
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
from keras.layers import Dense, Conv2D, MaxPooling2D, Flatten, BatchNormalization, Dropout
train dir = '/content/drive/MyDrive/Level 6/AI & ML/w6/FruitinAmazon/train'
img height = 224 # Example image height
img width = 224 # Example image width
batch size = 32
validation split = 0.2
rescale = tf.keras.layers.Rescaling(1./255) # Normalize pixel values to [0, 1]
train ds = tf.keras.preprocessing.image dataset from directory(
train dir,
labels='inferred',
label mode='int',
image size=(img height, img width),
interpolation='nearest',
batch size=batch size,
shuffle=True,
validation split=validation split,
subset='training',
seed=123
train_ds = train_ds.map(lambda x, y: (rescale(x), y))
Found 90 files belonging to 6 classes.
    Using 72 files for training.
```

```
# Create validation dataset with normalization
val_ds = tf.keras.preprocessing.image_dataset_from_directory(
train dir,
labels='inferred',
label mode='int',
image size=(img height, img width),
interpolation='nearest',
batch size=batch size,
shuffle=False,
validation split=validation split,
subset='validation',
seed=123
val ds = val ds.map(lambda x, y: (rescale(x), y))
Found 90 files belonging to 6 classes.
     Using 18 files for validation.
from tensorflow.keras.applications import VGG16
base model = VGG16(weights='imagenet', include top=False, input shape=(224, 224, 3))
base model.summary()
```



Layer (type)	Output Shape	Param #
input_layer (InputLayer)	(None, 224, 224, 3)	Ø
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1,792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36,928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73,856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147,584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295,168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590,080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590,080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1,180,160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0

```
Total params: 14,714,688 (56.13 MB)

for layer in base_model.layers:
    layer.trainable = False

model = Sequential()
model.add(base_model)
model.add(Flatten())
model.add(Dense(1024, activation='relu'))
model.add(Dense(6, activation='softmax'))
```

model.summary()

## **→** Mode:

## Model: "sequential"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14,714,688
flatten (Flatten)	(None, 25088)	0
dense (Dense)	(None, 1024)	25,691,136
dense_1 (Dense)	(None, 6)	6,150

Total params: 40,411,974 (154.16 MB)
Trainable params: 25,697,286 (98.03 MB)
Non-trainable params: 14.714.688 (56.13 MB)

model.compile(optimizer='adam', loss='sparse\_categorical\_crossentropy', metrics=['accuracy'])

```
history = None
with tf.device('/GPU:0'):
 print("GPU AVIALBLE")
 history = model.fit(train ds, validation data=val ds, epochs=10)
    GPU AVIALBLE
\rightarrow
     Epoch 1/10
     3/3 -----
                        Epoch 2/10
                          - 59s 19s/step - accuracy: 0.7383 - loss: 1.2681 - val_accuracy: 0.7778 - val_loss: 1.0839
     3/3 ---
    Epoch 3/10
     3/3 ----
                           - 81s 18s/step - accuracy: 0.9332 - loss: 0.2374 - val accuracy: 0.9444 - val loss: 0.3971
     Epoch 4/10
     3/3 -
                           - 82s 19s/step - accuracy: 0.8641 - loss: 0.5385 - val accuracy: 0.8889 - val loss: 0.5398
     Epoch 5/10
                          - 91s 23s/step - accuracy: 0.9705 - loss: 0.0740 - val accuracy: 0.8889 - val loss: 0.5773
     3/3 ----
     Epoch 6/10
                           - 71s 18s/step - accuracy: 0.9891 - loss: 0.0139 - val accuracy: 0.8889 - val loss: 0.6613
     3/3 -
     Epoch 7/10
                          — 58s 18s/step - accuracy: 1.0000 - loss: 0.0112 - val accuracy: 0.8889 - val loss: 0.6499
     3/3 ----
     Epoch 8/10
                          - 69s 25s/step - accuracy: 1.0000 - loss: 0.0060 - val accuracy: 0.8889 - val loss: 0.5778
     3/3 -
     Epoch 9/10
                           - 69s 18s/step - accuracy: 1.0000 - loss: 9.5737e-04 - val accuracy: 0.8889 - val loss: 0.5238
     3/3 ----
    Epoch 10/10
                          — 58s 18s/step - accuracy: 1.0000 - loss: 4.7257e-04 - val accuracy: 0.8889 - val loss: 0.4811
     3/3 ----
test loss, test acc = model.evaluate(val ds)
print(f"Validation Accuracy: {test acc:.2f}")
                       ---- 13s 13s/step - accuracy: 0.8889 - loss: 0.4811
```

Validation Accuracy: 0.89

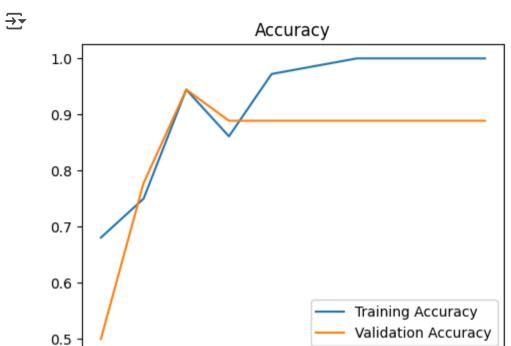
```
#plot for test data
# Plot training history
import matplotlib.pyplot as plt
plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy')
plt.legend()

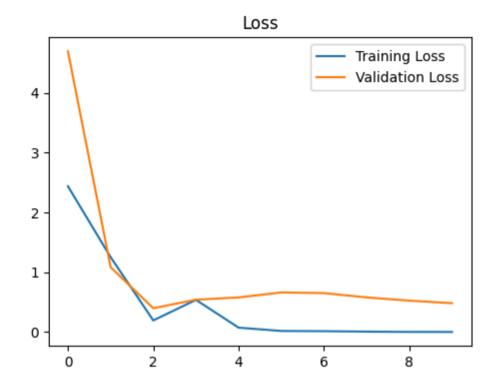
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss')
plt.legend()

plt.show()
```

4/10/25, 2:33 PM Worksheet6.ipynb - Colab



2



```
test_dir = "/content/drive/MyDrive/Level 6/AI & ML/w6/FruitinAmazon/test"
test_ds = tf.keras.preprocessing.image_dataset_from_directory(
    test_dir,
    labels='inferred',
    label_mode='int',
    image_size=(img_height, img_width),
    batch_size=batch_size,
    shuffle=False,
    interpolation='nearest',
    seed=123
)
test_ds = test_ds.map(lambda x, y: (rescale(x), y))
```

4

6

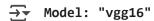
8

Found 30 files belonging to 6 classes. test loss, test accuracy = model.evaluate(test ds) print(f"Test Accuracy: {test\_accuracy:.4f}") print(f"Test Loss: {test loss:.4f}") 1/1 ----- 27s 27s/step - accuracy: 0.5667 - loss: 3.0391 Test Accuracy: 0.5667 Test Loss: 3.0391 import tensorflow from tensorflow import keras from keras import Sequential from keras.layers import Dense, Flatten from keras.applications.vgg16 import VGG16 from tensorflow.keras.optimizers import RMSprop base model = VGG16( weights='imagenet', include top = False, input shape=(224,224,3) base model.trainable = True set trainable = False for layer in base model.layers: if layer.name == 'block5 conv1': set trainable = True if set trainable: layer.trainable = True else: layer.trainable = False

for layer in base\_model.layers:
 print(layer.name,layer.trainable)

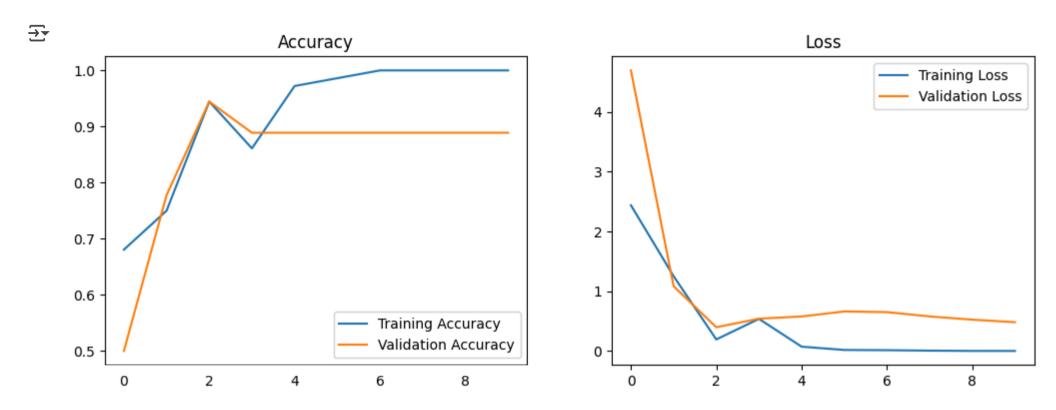
→ input\_layer\_2 False block1 conv1 False block1 conv2 False block1 pool False block2 conv1 False block2 conv2 False block2 pool False block3 conv1 False block3 conv2 False block3 conv3 False block3 pool False block4 conv1 False block4\_conv2 False block4 conv3 False block4 pool False block5 conv1 True block5 conv2 True block5 conv3 True block5 pool True

base\_model.summary()



Layer (type)	Output Shape	Param #
input_layer_2 (InputLayer)	(None, 224, 224, 3)	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1,792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36,928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73,856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147,584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295,168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590,080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590,080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1,180,160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2,359,808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2,359,808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0

```
Total params: 14,714,688 (56.13 MB)
model = Sequential()
model.add(base model)
model.add(Flatten())
model.add(Dense(256,activation='relu'))
model.add(Dense(6,activation='softmax'))
model.compile(optimizer=keras.optimizers.RMSprop(learning rate=1e-5), loss='sparse categorical crossentropy', metrics=['accuracy'])
test loss, test acc = model.evaluate(val ds)
print(f"Validation Accuracy: {test acc:.2f}")
Validation Accuracy: 0.17
#plot for test data
# Plot training history
import matplotlib.pyplot as plt
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val accuracy'], label='Validation Accuracy'
plt.title('Accuracy')
plt.legend()
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val loss'], label='Validation Loss')
plt.title('Loss')
plt.legend()
```



test\_dir = "/content/drive/MyDrive/Level 6/AI & ML/w6/FruitinAmazon/test"

test ds = tf.keras.preprocessing.image dataset from directory(