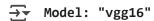
```
import tensorflow as tf
from tensorflow import keras
img_height =224
img_width =224
batch_size = 32
validation_split = 0.2
rescale = tf.keras.layers.Rescaling(1./255)
import zipfile
zip_ref = zipfile.ZipFile('/content/drive/MyDrive/Level 6/Artificial_Intelligence/Week5/FruitinAmazon.zip', 'r')
zip_ref.extractall('/content')
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/Artificial Intelligence/Week5/FruitinAmazon/train'
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))
     Downloading data from <a href="https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16">https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16</a> weights tf dim ordering
      58889256/58889256 -
                                               - 0s 0us/step
base_model.summary()
```

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Layer (type)	Output Shape	Param #
input_layer (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)

```
layer.trainable = False

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

for layer in base_model.layers:

model.summary()

→ 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 = model.fit(train ds, validation data=val ds, epochs=10)

```
\rightarrow Epoch 1/10
                              73s 19s/step - accuracy: 0.1997 - loss: 7.5966 - val_accuracy: 0.1111 - val_loss: 10.7234
    3/3 ----
    Epoch 2/10
                            - 73s 20s/step - accuracy: 0.4913 - loss: 8.9173 - val accuracy: 0.8333 - val loss: 2.3984
    3/3 -
    Epoch 3/10
                            - 50s 16s/step - accuracy: 0.6289 - loss: 8.3381 - val accuracy: 0.8333 - val loss: 0.9623
    3/3 ----
    Epoch 4/10
                              82s 16s/step - accuracy: 0.7700 - loss: 5.0412 - val accuracy: 0.6111 - val loss: 3.2832
    3/3 ----
    Epoch 5/10
                              61s 21s/step - accuracy: 0.7639 - loss: 2.5738 - val accuracy: 0.6111 - val loss: 3.5414
    3/3 ----
    Epoch 6/10
    3/3 ----
                              82s 22s/step - accuracy: 0.9449 - loss: 0.3297 - val accuracy: 0.4444 - val loss: 3.0220
    Epoch 7/10
                              50s 16s/step - accuracy: 0.8806 - loss: 0.4283 - val accuracy: 0.7778 - val loss: 1.0958
    3/3 -
    Epoch 8/10
                            - 51s 17s/step - accuracy: 0.9627 - loss: 0.0748 - val accuracy: 0.8889 - val loss: 0.4856
    3/3 ----
    Epoch 9/10
                            — 61s 21s/step - accuracy: 1.0000 - loss: 0.0049 - val accuracy: 0.9444 - val loss: 0.2508
    3/3 ----
    Epoch 10/10
                            - 55s 17s/step - accuracy: 1.0000 - loss: 0.0177 - val accuracy: 0.9444 - val loss: 0.2098
    3/3 ----
```

test_loss, test_acc = model.evaluate(val_ds)
print(f"Validation Accuracy: {test_acc:.2f}")

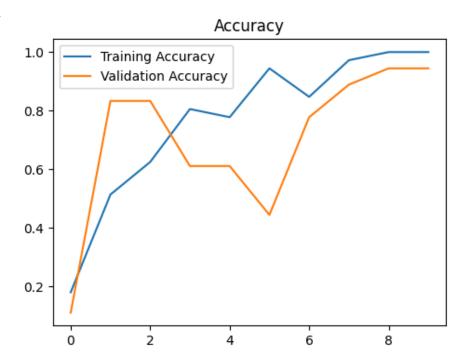
```
#plot for test data
# Plot training history
import matplotlib.pyplot as plt
plt.figure(figsize=(12, 4))

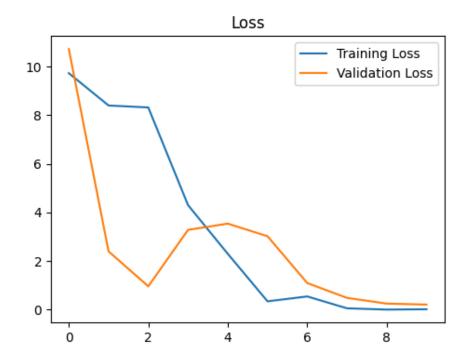
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy')
plt.legend()

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()
```







```
test_dir = "/content/drive/MyDrive/Level 6/Artificial_Intelligence/Week5/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
)

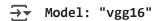
test_ds = test_ds.map(lambda x, y: (rescale(x), y))

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 — 20s 20s/step - accuracy: 0.2667 - loss: 6.7009
     Test Accuracy: 0.2667
     Test Loss: 6.7009
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=['acc
test loss, test acc = model.evaluate(val ds)
print(f"Validation Accuracy: {test_acc:.2f}")
Validation Accuracy: 0.00
#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()
```



