Task 1

```
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
   Improve the Previous Model
from \ tensorflow.keras.preprocessing.image \ import \ ImageDataGenerator
train_datagen = ImageDataGenerator(
   rescale=1./255,
   rotation range=20,
   width_shift_range=0.2,
   height_shift_range=0.2,
   shear_range=0.2,
   zoom_range=0.2,
   horizontal_flip=True,
   fill_mode='nearest'
)
test_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow_from_directory(
    '/content/drive/MyDrive/AlandML/Worksheet5/FruitinAmazon/FruitinAmazon/train',
   target_size=(224, 224),
   batch_size=32,
   class_mode='categorical'
validation_generator = test_datagen.flow_from_directory(
    '/content/drive/MyDrive/AIandML/Worksheet5/FruitinAmazon/FruitinAmazon/test',
   target_size=(224, 224),
   batch_size=32,
   class_mode='categorical'
)
    Found 90 images belonging to 6 classes.
     Found 30 images belonging to 6 classes.
```

Build a Deeper CNN Model with Batch Normalization & Dropout

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormalization
model = Sequential([
   Conv2D(32, (3,3), activation='relu', input_shape=(64, 64, 3)),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2,2)),
   Conv2D(64, (3,3), activation='relu'),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2,2)),
   Dropout(0.3),
   Conv2D(128, (3,3), activation='relu'),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2,2)),
   Dropout(0.4),
   Flatten(),
   Dense(256, activation='relu'),
   Dropout(0.5),
   Dense(10, activation='softmax')
])
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

//wsr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`inp super().__init__(activity_regularizer=activity_regularizer, **kwargs)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 32)	896
batch_normalization (BatchNormalization)	(None, 62, 62, 32)	128
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
conv2d_1 (Conv2D)	(None, 29, 29, 64)	18,496
batch_normalization_1 (BatchNormalization)	(None, 29, 29, 64)	256
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout (Dropout)	(None, 14, 14, 64)	0
conv2d_2 (Conv2D)	(None, 12, 12, 128)	73,856
batch_normalization_2 (BatchNormalization)	(None, 12, 12, 128)	512
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 128)	0
dropout_1 (Dropout)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 256)	1,179,904
dropout_2 (Dropout)	(None, 256)	0
dense_1 (Dense)	(None, 10)	2,570

Total params: 1,276,618 (4.87 MB)

Transfer Learning with VGG16

```
from tensorflow.keras.applications import VGG16
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dense, GlobalAveragePooling2D

# Load VGG16 without the top classification layer
base_model = VGG16(weights='imagenet', include_top=False, input_shape=(224, 224, 3))

# Freeze base model layers
```

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

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16 weights tf dim ordering tf kernels notop 58889256/58889256 ———— Os Ous/step

Add Custom Layers

```
num_classes = len(train_generator.class_indices)
x = base_model.output
x = GlobalAveragePooling2D()(x)
x = Dense(1024, activation='relu')(x)
x = Dense(num_classes, activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=x)
```

Compile & Train the Model

```
from tensorflow.keras.optimizers import Adam
model.compile(optimizer=Adam(), loss='categorical_crossentropy', metrics=['accuracy'])
history = model.fit(train_generator, epochs=10, validation_data=validation_generator)
```

```
🚁 /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class
      self._warn_if_super_not_called()
    Epoch 1/10
    3/3 -
                            - 54s 18s/step - accuracy: 0.1865 - loss: 1.8096 - val_accuracy: 0.4000 - val_loss: 1.6733
    Epoch 2/10
    3/3 ·
                            - 31s 605ms/step - accuracy: 0.3470 - loss: 1.6162 - val_accuracy: 0.4667 - val_loss: 1.5452
    Epoch 3/10
                            - 2s 578ms/step - accuracy: 0.6265 - loss: 1.4294 - val_accuracy: 0.4333 - val_loss: 1.5507
    3/3
    Epoch 4/10
    3/3
                            - 2s 660ms/step - accuracy: 0.6060 - loss: 1.2728 - val_accuracy: 0.3667 - val_loss: 1.4750
    Epoch 5/10
    3/3
                            - 2s 578ms/step - accuracy: 0.6135 - loss: 1.1824 - val_accuracy: 0.6667 - val_loss: 1.3536
    Epoch 6/10
                            - 2s 600ms/step - accuracy: 0.7307 - loss: 1.1001 - val_accuracy: 0.5333 - val_loss: 1.3242
    3/3
    Epoch 7/10
                            • 2s 610ms/step - accuracy: 0.7319 - loss: 0.9823 - val_accuracy: 0.4000 - val_loss: 1.3343
    3/3
    Epoch 8/10
    3/3
                             2s 794ms/step - accuracy: 0.7589 - loss: 0.9183 - val_accuracy: 0.4667 - val_loss: 1.3425
    Epoch 9/10
                            - 2s 598ms/step - accuracy: 0.8035 - loss: 0.8774 - val accuracy: 0.6333 - val loss: 1.2571
    3/3
    Epoch 10/10
                             2s 634ms/step - accuracy: 0.8953 - loss: 0.7401 - val_accuracy: 0.6000 - val_loss: 1.2084
```

Model Analysis

model.summary()

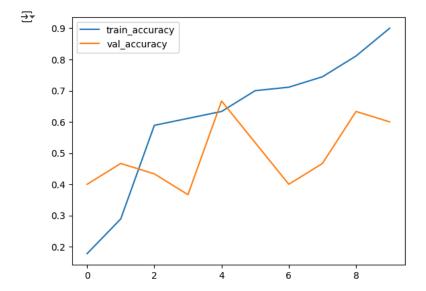
→ Model: "functional_1"

Layer (type)	Output Shape	Param #
input_layer_1 (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
global_average_pooling2d (GlobalAveragePooling2D)	(None, 512)	0
dense_2 (Dense)	(None, 1024)	525,312
dense_3 (Dense)	(None, 6)	6,150

Total params: 16,309,076 (62.21 MB)

Visualize Training Performance

```
import matplotlib.pyplot as plt
# history = model.fit(train_generator, epochs=10, validation_data=validation_generator)
plt.plot(history.history['accuracy'], label='train_accuracy')
plt.plot(history.history['val_accuracy'], label='val_accuracy')
plt.legend()
plt.show()
```



✓ Task 2

Data Preparation & Augmentation

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
train_dir = '/content/drive/MyDrive/AlandML/Worksheet5/FruitinAmazon/FruitinAmazon/train'
val_dir = '/content/drive/MyDrive/AIandML/Worksheet5/FruitinAmazon/FruitinAmazon/test'
# Data augmentation
train_datagen = ImageDataGenerator(
    rescale=1./255,
    rotation_range=20,
    zoom range=0.2,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True
)
val_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(224, 224),
    batch_size=32,
    class_mode='categorical'
validation_generator = val_datagen.flow_from_directory(
    val_dir,
    target_size=(224, 224),
    batch_size=32,
    class_mode='categorical',
    shuffle=False # Important for inference output
)
num_classes = len(train_generator.class_indices)
     Found 90 images belonging to 6 classes.
     Found 30 images belonging to 6 classes.
```

Load VGG16 (Pre-trained) and Freeze Layers

```
from tensorflow.keras.applications import VGG16
from tensorflow.keras.models import Model
from tensorflow.keras.layers import GlobalAveragePooling2D, Dense, Dropout
from tensorflow.keras.optimizers import Adam
```

```
# Freeze all layers
for layer in base_model.layers:
    layer.trainable = False
Add Custom Layers
x = base_model.output
x = GlobalAveragePooling2D()(x)
x = Dense(512, activation='relu')(x)
x = Dropout(0.5)(x)
x = Dense(num_classes, activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=x)
   Compile & Train the Model
model.compile(optimizer=Adam(learning_rate=0.0001), loss='categorical_crossentropy', metrics=['accuracy'])
history = model.fit(
    train_generator,
    epochs=10.
    validation_data=validation_generator
)
→ Epoch 1/10
                            — 6s 1s/step - accuracy: 0.1293 - loss: 2.1646 - val_accuracy: 0.1667 - val_loss: 1.9615
     3/3 -
     Epoch 2/10
     3/3 -
                            - 2s 578ms/step - accuracy: 0.1104 - loss: 2.0926 - val_accuracy: 0.1333 - val_loss: 1.8992
     Epoch 3/10
     3/3 -
                            - 2s 704ms/step - accuracy: 0.1158 - loss: 1.9014 - val_accuracy: 0.1000 - val_loss: 1.8569
     Epoch 4/10
                            - 3s 792ms/step - accuracy: 0.2099 - loss: 1.8784 - val_accuracy: 0.0667 - val_loss: 1.8298
     3/3 -
     Epoch 5/10
                            - 2s 650ms/step - accuracy: 0.1573 - loss: 1.8747 - val accuracy: 0.1333 - val loss: 1.8111
     3/3 -
     Epoch 6/10
                            - 2s 586ms/step - accuracy: 0.1787 - loss: 1.7941 - val_accuracy: 0.2000 - val_loss: 1.7974
     3/3 .
     Epoch 7/10
     3/3
                            - 2s 585ms/step - accuracy: 0.1425 - loss: 1.8980 - val_accuracy: 0.2333 - val_loss: 1.7872
     Epoch 8/10
     3/3 -
                            - 2s 580ms/step - accuracy: 0.1748 - loss: 1.8456 - val_accuracy: 0.2333 - val_loss: 1.7791
     Epoch 9/10
                            - 2s 583ms/step - accuracy: 0.1832 - loss: 1.8352 - val_accuracy: 0.2000 - val_loss: 1.7717
     3/3 -
     Epoch 10/10
                            - 2s 783ms/step - accuracy: 0.2802 - loss: 1.8221 - val_accuracy: 0.3000 - val_loss: 1.7646
  Evaluate + Classification Report
from sklearn.metrics import classification_report
import numpy as np
# Predict on validation data
y_pred = model.predict(validation_generator)
```

base model = VGG16(weights='imagenet', include top=False, input shape=(224, 224, 3))