```
In [2]:
         import tensorflow as tf
         from tensorflow.keras.applications import EfficientNetB0
         from tensorflow.keras.preprocessing.image import ImageDataGenerator
         from tensorflow.keras.models import Sequential
         from tensorflow.keras.layers import Dense, Dropout, BatchNormalization
         from tensorflow.keras.optimizers import Adam
         from tensorflow.keras.losses import CategoricalCrossentropy
         from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
         # Directories where your data is stored
         train_dir = r'C:\Users\Abhishek\Downloads\cucumber\train'
         validation_dir = r'C:\Users\Abhishek\Downloads\cucumber\cucumber\valid'
         test_dir = r'C:\Users\Abhishek\Downloads\cucumber\cucumber\test'
         # Define constants
         IMG SIZE = 224
         BATCH_SIZE = 32 # Reduced batch size
         NUM_CLASSES = 8 # Number of classes in your dataset
         EPOCHS = 100
         # Generate batches of tensor image data with real-time data augmentation
         datagen = ImageDataGenerator(
             rescale=1./255,
             horizontal_flip=True,
             vertical_flip=True)
         train_generator = datagen.flow_from_directory(
             train_dir,
             target_size=(IMG_SIZE, IMG_SIZE),
             batch_size=BATCH_SIZE,
             class mode='categorical')
         validation_generator = datagen.flow_from_directory(
             validation_dir,
             target size=(IMG_SIZE, IMG_SIZE),
             batch_size=BATCH_SIZE,
             class_mode='categorical')
         test_generator = datagen.flow_from_directory(
             test dir,
             target_size=(IMG_SIZE, IMG_SIZE),
             batch_size=BATCH_SIZE,
             class mode='categorical')
         # Load base model
         base_model = EfficientNetB0(weights='imagenet', include_top=False, input_shape=(IMG_
         # Add a new top Layer
         x = base model.output
         x = tf.keras.layers.GlobalAveragePooling2D()(x)
         x = Dense(512, activation='relu')(x) # Reduced the number of neurons
         x = Dropout(0.2)(x) # Add dropout layer to reduce overfitting
         x = BatchNormalization()(x)
         predictions = Dense(NUM_CLASSES, activation='softmax')(x)
         # This is the model we will train
         model = tf.keras.models.Model(inputs=base model.input, outputs=predictions)
         # Freeze the base model
         for layer in base model.layers:
             layer.trainable = False
```

```
# Compile the model
model.compile(optimizer=Adam(lr=0.001), loss=CategoricalCrossentropy(), metrics=['ac
# Define callbacks
early_stopping = EarlyStopping(monitor='val_loss', patience=10, restore_best_weights
reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.2, patience=5, min_lr=0.0
# Train the model
history = model.fit(
   train_generator,
   epochs=EPOCHS,
   validation_data=validation_generator,
   callbacks=[early_stopping, reduce_lr])
# Unfreeze the layers of the base model and fine-tune the entire model
for layer in base model.layers:
   layer.trainable = True
# Recompile the model
model.compile(optimizer=Adam(lr=0.00001), loss=CategoricalCrossentropy(), metrics=['
# Continue training the model
history_fine_tuning = model.fit(
   train_generator,
   epochs=EPOCHS,
   validation data=validation generator,
   callbacks=[early_stopping, reduce_lr]) b
# Evaluate the model on the test data after fine-tuning
# Evaluate the model on the test data after fine-tuning
score = model.evaluate(test generator)
print(f'Test loss: {score[0]} / Test accuracy: {score[1]}')g
Found 800 images belonging to 8 classes.
Found 240 images belonging to 8 classes.
Found 240 images belonging to 8 classes.
WARNING:absl:`lr` is deprecated, please use `learning_rate` instead, or use the lega
cy optimizer, e.g., tf.keras.optimizers.legacy.Adam.
Epoch 1/100
175 - val_loss: 2.3123 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 2/100
225 - val_loss: 2.3302 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 3/100
300 - val_loss: 2.2826 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 4/100
425 - val_loss: 2.2736 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 5/100
225 - val_loss: 2.2586 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 6/100
00 - val_loss: 2.3918 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 7/100
75 - val loss: 2.3011 - val accuracy: 0.1250 - lr: 0.0010
Epoch 8/100
00 - val_loss: 2.1480 - val_accuracy: 0.1375 - lr: 0.0010
Epoch 9/100
12 - val_loss: 2.1358 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 10/100
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262 - val_loss: 2.1200 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 11/100
25/25 [================= ] - 259s 10s/step - loss: 2.1673 - accuracy: 0.
1688 - val_loss: 2.1211 - val_accuracy: 0.1292 - lr: 0.0010
Epoch 12/100
25/25 [================= ] - 275s 11s/step - loss: 2.1992 - accuracy: 0.
1350 - val_loss: 2.0963 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 13/100
25/25 [================== ] - 280s 11s/step - loss: 2.1851 - accuracy: 0.
1088 - val_loss: 2.0892 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 14/100
25/25 [================== ] - 282s 11s/step - loss: 2.1678 - accuracy: 0.
1163 - val loss: 2.0961 - val accuracy: 0.1250 - lr: 0.0010
Epoch 15/100
25/25 [============== ] - 280s 11s/step - loss: 2.1536 - accuracy: 0.
1325 - val loss: 2.0858 - val accuracy: 0.1250 - lr: 0.0010
Epoch 16/100
25/25 [================= ] - 338s 13s/step - loss: 2.1510 - accuracy: 0.
1325 - val loss: 2.0963 - val accuracy: 0.1250 - lr: 0.0010
Epoch 17/100
25/25 [============== ] - 365s 15s/step - loss: 2.1193 - accuracy: 0.
1425 - val_loss: 2.0923 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 18/100
1275 - val_loss: 2.1216 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 19/100
25/25 [================== ] - 378s 15s/step - loss: 2.1244 - accuracy: 0.
1262 - val_loss: 2.0860 - val_accuracy: 0.1750 - lr: 0.0010
Epoch 20/100
25/25 [================= ] - 377s 15s/step - loss: 2.1274 - accuracy: 0.
1325 - val_loss: 2.1012 - val_accuracy: 0.1708 - lr: 0.0010
Epoch 21/100
25/25 [================= ] - 378s 15s/step - loss: 2.1263 - accuracy: 0.
1338 - val_loss: 2.0902 - val_accuracy: 0.1250 - lr: 2.0000e-04
Epoch 22/100
25/25 [================= ] - 280s 11s/step - loss: 2.1279 - accuracy: 0.
1375 - val_loss: 2.0874 - val_accuracy: 0.1250 - lr: 2.0000e-04
Epoch 23/100
363 - val_loss: 2.0853 - val_accuracy: 0.1250 - lr: 2.0000e-04
Epoch 24/100
238 - val_loss: 2.0797 - val_accuracy: 0.1958 - lr: 2.0000e-04
Epoch 25/100
388 - val loss: 2.0832 - val accuracy: 0.1708 - lr: 2.0000e-04
Epoch 26/100
200 - val loss: 2.1433 - val accuracy: 0.1250 - lr: 2.0000e-04
Epoch 27/100
200 - val loss: 2.1013 - val accuracy: 0.1500 - lr: 2.0000e-04
Epoch 28/100
488 - val loss: 2.0869 - val accuracy: 0.1375 - lr: 2.0000e-04
Epoch 29/100
138 - val loss: 2.0818 - val accuracy: 0.1417 - lr: 2.0000e-04
Epoch 30/100
462 - val loss: 2.0788 - val accuracy: 0.1375 - lr: 4.0000e-05
Epoch 31/100
637 - val loss: 2.0771 - val accuracy: 0.1292 - lr: 4.0000e-05
Epoch 32/100
412 - val loss: 2.0760 - val accuracy: 0.1250 - lr: 4.0000e-05
Epoch 33/100
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25/25 [===========] - 208s 8s/step - loss: 2.1238 - accuracy: 0.1
338 - val_loss: 2.0752 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 34/100
450 - val_loss: 2.0734 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 35/100
300 - val_loss: 2.0727 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 36/100
163 - val_loss: 2.0731 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 37/100
388 - val loss: 2.0730 - val accuracy: 0.1250 - lr: 4.0000e-05
Epoch 38/100
475 - val loss: 2.0707 - val accuracy: 0.1250 - lr: 4.0000e-05
Epoch 39/100
312 - val loss: 2.0697 - val accuracy: 0.1250 - lr: 4.0000e-05
Epoch 40/100
538 - val_loss: 2.0686 - val_accuracy: 0.1375 - lr: 4.0000e-05
Epoch 41/100
213 - val_loss: 2.0677 - val_accuracy: 0.2417 - lr: 4.0000e-05
Epoch 42/100
388 - val_loss: 2.0675 - val_accuracy: 0.1375 - lr: 4.0000e-05
Epoch 43/100
262 - val_loss: 2.0674 - val_accuracy: 0.1750 - lr: 4.0000e-05
Epoch 44/100
213 - val_loss: 2.0676 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 45/100
150 - val_loss: 2.0664 - val_accuracy: 0.1500 - lr: 4.0000e-05
Epoch 46/100
325 - val_loss: 2.0647 - val_accuracy: 0.1917 - lr: 4.0000e-05
Epoch 47/100
338 - val_loss: 2.0654 - val_accuracy: 0.1333 - lr: 4.0000e-05
Epoch 48/100
275 - val loss: 2.0654 - val accuracy: 0.1250 - lr: 4.0000e-05
Epoch 49/100
513 - val loss: 2.0641 - val accuracy: 0.1625 - lr: 4.0000e-05
Epoch 50/100
300 - val loss: 2.0642 - val accuracy: 0.2667 - lr: 4.0000e-05
Epoch 51/100
475 - val loss: 2.0645 - val accuracy: 0.1708 - lr: 4.0000e-05
Epoch 52/100
275 - val loss: 2.0639 - val accuracy: 0.2000 - lr: 4.0000e-05
Epoch 53/100
350 - val loss: 2.0618 - val accuracy: 0.2250 - lr: 4.0000e-05
Epoch 54/100
1450 - val_loss: 2.0651 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 55/100
425 - val loss: 2.0605 - val accuracy: 0.2500 - lr: 4.0000e-05
Epoch 56/100
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25/25 [===========] - 228s 9s/step - loss: 2.1020 - accuracy: 0.1
550 - val_loss: 2.0621 - val_accuracy: 0.1833 - lr: 4.0000e-05
Epoch 57/100
575 - val_loss: 2.0612 - val_accuracy: 0.2000 - lr: 4.0000e-05
Epoch 58/100
538 - val_loss: 2.0608 - val_accuracy: 0.1708 - lr: 4.0000e-05
Epoch 59/100
425 - val_loss: 2.0602 - val_accuracy: 0.1417 - lr: 4.0000e-05
Epoch 60/100
538 - val loss: 2.0602 - val accuracy: 0.2000 - lr: 4.0000e-05
Epoch 61/100
475 - val loss: 2.0582 - val accuracy: 0.2042 - lr: 4.0000e-05
Epoch 62/100
363 - val loss: 2.0592 - val accuracy: 0.1625 - lr: 4.0000e-05
Epoch 63/100
475 - val_loss: 2.0603 - val_accuracy: 0.1333 - lr: 4.0000e-05
Epoch 64/100
513 - val_loss: 2.0637 - val_accuracy: 0.1250 - lr: 4.0000e-05
Epoch 65/100
538 - val_loss: 2.0608 - val_accuracy: 0.1417 - lr: 4.0000e-05
Epoch 66/100
450 - val_loss: 2.0600 - val_accuracy: 0.1792 - lr: 4.0000e-05
Epoch 67/100
312 - val_loss: 2.0589 - val_accuracy: 0.1417 - lr: 1.0000e-05
Epoch 68/100
412 - val_loss: 2.0579 - val_accuracy: 0.1417 - lr: 1.0000e-05
Epoch 69/100
612 - val_loss: 2.0570 - val_accuracy: 0.1667 - lr: 1.0000e-05
Epoch 70/100
462 - val_loss: 2.0573 - val_accuracy: 0.1792 - lr: 1.0000e-05
Epoch 71/100
450 - val loss: 2.0575 - val accuracy: 0.1333 - lr: 1.0000e-05
Epoch 72/100
525 - val loss: 2.0562 - val accuracy: 0.1708 - lr: 1.0000e-05
Epoch 73/100
488 - val loss: 2.0556 - val accuracy: 0.1500 - lr: 1.0000e-05
Epoch 74/100
462 - val loss: 2.0541 - val accuracy: 0.2042 - lr: 1.0000e-05
Epoch 75/100
500 - val loss: 2.0536 - val accuracy: 0.1833 - lr: 1.0000e-05
Epoch 76/100
450 - val loss: 2.0540 - val accuracy: 0.1792 - lr: 1.0000e-05
Epoch 77/100
688 - val_loss: 2.0551 - val_accuracy: 0.1625 - lr: 1.0000e-05
Epoch 78/100
575 - val loss: 2.0546 - val accuracy: 0.1750 - lr: 1.0000e-05
Epoch 79/100
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25/25 [===========] - 224s 9s/step - loss: 2.0845 - accuracy: 0.1
538 - val_loss: 2.0528 - val_accuracy: 0.1750 - lr: 1.0000e-05
Epoch 80/100
450 - val_loss: 2.0533 - val_accuracy: 0.1792 - lr: 1.0000e-05
Epoch 81/100
375 - val_loss: 2.0545 - val_accuracy: 0.1375 - lr: 1.0000e-05
Epoch 82/100
713 - val_loss: 2.0530 - val_accuracy: 0.1417 - lr: 1.0000e-05
Epoch 83/100
462 - val loss: 2.0514 - val accuracy: 0.1708 - lr: 1.0000e-05
Epoch 84/100
538 - val_loss: 2.0526 - val_accuracy: 0.1417 - lr: 1.0000e-05
Epoch 85/100
388 - val_loss: 2.0528 - val_accuracy: 0.1542 - lr: 1.0000e-05
Epoch 86/100
725 - val_loss: 2.0529 - val_accuracy: 0.1417 - lr: 1.0000e-05
Epoch 87/100
775 - val_loss: 2.0526 - val_accuracy: 0.1708 - lr: 1.0000e-05
Epoch 88/100
488 - val_loss: 2.0510 - val_accuracy: 0.1750 - lr: 1.0000e-05
Epoch 89/100
625 - val_loss: 2.0486 - val_accuracy: 0.1792 - lr: 1.0000e-05
Epoch 90/100
450 - val_loss: 2.0502 - val_accuracy: 0.1750 - lr: 1.0000e-05
Epoch 91/100
513 - val_loss: 2.0495 - val_accuracy: 0.2042 - lr: 1.0000e-05
Epoch 92/100
338 - val_loss: 2.0482 - val_accuracy: 0.2167 - lr: 1.0000e-05
Epoch 93/100
800 - val_loss: 2.0485 - val_accuracy: 0.2042 - lr: 1.0000e-05
Epoch 94/100
575 - val loss: 2.0509 - val accuracy: 0.1917 - lr: 1.0000e-05
Epoch 95/100
462 - val loss: 2.0483 - val accuracy: 0.1875 - lr: 1.0000e-05
Epoch 96/100
675 - val loss: 2.0490 - val accuracy: 0.1750 - lr: 1.0000e-05
Epoch 97/100
437 - val loss: 2.0471 - val accuracy: 0.2000 - lr: 1.0000e-05
Epoch 98/100
513 - val loss: 2.0469 - val accuracy: 0.2042 - lr: 1.0000e-05
Epoch 99/100
800 - val loss: 2.0485 - val accuracy: 0.2000 - lr: 1.0000e-05
Epoch 100/100
587 - val loss: 2.0498 - val accuracy: 0.2083 - lr: 1.0000e-05
WARNING:absl:`lr` is deprecated, please use `learning_rate` instead, or use the lega
cy optimizer, e.g.,tf.keras.optimizers.legacy.Adam.
Epoch 1/100
25/25 [================ ] - 405s 13s/step - loss: 0.7465 - accuracy: 0.
```

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7300 - val_loss: 2.4014 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 2/100
9212 - val_loss: 2.4741 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 3/100
9287 - val_loss: 3.0748 - val_accuracy: 0.1250 - lr: 0.0010
Epoch 4/100
25/25 [================ ] - 359s 14s/step - loss: 0.2363 - accuracy: 0.
9287 - val_loss: 50.9114 - val_accuracy: 0.1000 - lr: 0.0010
Epoch 5/100
25/25 [=============== ] - 330s 13s/step - loss: 0.1445 - accuracy: 0.
9600 - val_loss: 66.9357 - val_accuracy: 0.1792 - lr: 0.0010
Epoch 6/100
25/25 [============== ] - 332s 13s/step - loss: 0.1097 - accuracy: 0.
9625 - val loss: 13.8884 - val accuracy: 0.1417 - lr: 0.0010
Epoch 7/100
25/25 [============= ] - 329s 13s/step - loss: 0.0627 - accuracy: 0.
9762 - val loss: 11.0382 - val accuracy: 0.1417 - lr: 2.0000e-04
Epoch 8/100
25/25 [============== ] - 327s 13s/step - loss: 0.0459 - accuracy: 0.
9862 - val loss: 7.5471 - val accuracy: 0.1708 - lr: 2.0000e-04
Epoch 9/100
25/25 [============= ] - 336s 13s/step - loss: 0.0157 - accuracy: 0.
9975 - val_loss: 7.7141 - val_accuracy: 0.1208 - lr: 2.0000e-04
Epoch 10/100
25/25 [=============== ] - 325s 13s/step - loss: 0.0357 - accuracy: 0.
9912 - val_loss: 17.1959 - val_accuracy: 0.0917 - lr: 2.0000e-04
Epoch 11/100
25/25 [================ ] - 329s 13s/step - loss: 0.0266 - accuracy: 0.
9937 - val_loss: 18.3053 - val_accuracy: 0.1083 - lr: 2.0000e-04
Test loss: 2.4016289710998535 / Test accuracy: 0.125
```

In [ ]: