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Bokeh

Klasifikasi Jenis Buah Anggur (Merah, Thompson, Concord)

MobileNet

```
import os
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential, load_model
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dense,
Dropout, Flatten

count = 0
dirs = os.listdir(r'C:\Users\lenovo\OneDrive\Documents\TubesPMDPM\
train_data')
for dir in dirs:
    files = list(os.listdir(r'C:\Users\lenovo\OneDrive\Documents\
TubesPMDPM\train_data/'+dir))
    print(dir + ' Folder has ' + str(len(files)) + ' Images')
    count = count + len(files)
print('Images Folder has ' + str(count) + ' Images')

Concord Folder has 100 Images
Merah Folder has 110 Images
Thompson Folder has 100 Images
Images Folder has 310 Images

base_dir = r'C:\Users\lenovo\OneDrive\Documents\TubesPMDPM\train_data'
img_size = 180
batch = 16
validation_split = 0.1

num_classes = 3

dataset = tf.keras.utils.image_dataset_from_directory(
    base_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
    class_names=['Merah', 'Thompson', 'Concord']
)

Found 310 files belonging to 3 classes.
```

```

class_names = dataset.class_names
print("Class Names:", class_names)

expected_classes = ['Merah', 'Thompson', 'Concord']
if set(class_names) != set(expected_classes):
    print("Warning: Kelas yang diambil tidak sesuai dengan yang diharapkan!")

Class Names: ['Merah', 'Thompson', 'Concord']

```

Pembagian Dataset menjadi Train, Validation, dan Test dengan rasio 80:10:10

```

total_count = len(dataset)
train_count = int(total_count * 0.8)
val_count = int(total_count * 0.1)
test_count = total_count - train_count - val_count

if train_count < 0 or val_count < 0 or test_count < 0:
    raise ValueError("Jumlah gambar untuk pelatihan, validasi, atau pengujian tidak valid!")

print("Total Images:", total_count)
print("Train Images:", train_count)
print("Validation Images:", val_count)
print("Test Images:", test_count)

```

```

Total Images: 20
Train Images: 16
Validation Images: 2
Test Images: 2

```

```

train_ds = dataset.take(train_count)
val_ds = dataset.skip(train_count)

print(f"Dataset telah dibagi menjadi {train_count} gambar untuk pelatihan dan {val_count} gambar untuk validasi.")

Dataset telah dibagi menjadi 16 gambar untuk pelatihan dan 2 gambar untuk validasi.

```

```

import matplotlib.pyplot as plt

plt.figure(figsize=(10,10))

for images, labels in train_ds.take(1):
    num_images_to_display = min(9, len(images))
    for i in range(num_images_to_display):
        plt.subplot(3, 3, i + 1)
        plt.imshow(images[i].numpy().astype('uint8'))
        plt.title(class_names[labels[i]])

```

```
plt.axis('off')  
plt.show()
```

Thompson



Thompson



Thompson



Thompson



Thompson



Merah



Concord



Concord



Concord



```
import numpy as np  
  
for images, labels in train_ds.take(1):  
    images_array = np.array(images)  
    print(images_array.shape)  
  
(16, 180, 180, 3)
```

```

AUTOTUNE = tf.data.AUTOTUNE

train_ds = train_ds.cache().shuffle(1000).prefetch(buffer_size =
AUTOTUNE)
val_ds = val_ds.cache().shuffle(1000).prefetch(buffer_size = AUTOTUNE)

data_augmentation = Sequential([
    layers.RandomFlip("horizontal", input_shape=(img_size, img_size,
3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1),
    layers.RandomBrightness(0.1)
])

plt.figure(figsize=(10, 10))

for images, labels in train_ds.take(1):
    num_images_to_display = min(9, len(images))
    for i in range(num_images_to_display):
        augmented_image = data_augmentation(images[i:i+1])
        plt.subplot(3, 3, i + 1)
        plt.imshow(augmented_image[0].numpy().astype('uint8'))
        plt.axis('off')

plt.show()

c:\Users\lenovo\anaconda3\Lib\site-packages\keras\src\layers\
preprocessing\tf_data_layer.py:19: UserWarning: Do not pass an
`input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in
the model instead.
  super().__init__(**kwargs)

```




```
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential

# Membuat model dari awal
model = Sequential([
    layers.Rescaling(1./255, input_shape=(img_size, img_size, 3)),
    layers.Conv2D(32, (3, 3), activation='relu'),
    layers.BatchNormalization(),
    layers.MaxPooling2D(pool_size=(2, 2)),

    layers.Conv2D(64, (3, 3), activation='relu'),
```

```

layers.BatchNormalization(),
layers.MaxPooling2D(pool_size=(2, 2)),

layers.Conv2D(128, (3, 3), activation='relu'),
layers.BatchNormalization(),
layers.MaxPooling2D(pool_size=(2, 2)),

layers.Conv2D(256, (3, 3), activation='relu'),
layers.BatchNormalization(),
layers.MaxPooling2D(pool_size=(2, 2)),

layers.GlobalAveragePooling2D(),
layers.Dense(128, activation='relu'),
layers.Dropout(0.3),
layers.Dense(len(class_names), activation='softmax')
])

```

```

from tensorflow.keras.optimizers import Adam

```

```

model.compile(
    optimizer=Adam(learning_rate=1e-4),
    loss='sparse_categorical_crossentropy',
    metrics=['accuracy']
)

```

```

model.summary()

```

Model: "sequential_3"

Layer (type) Param #	Output Shape	
rescaling_1 (Rescaling)	(None, 180, 180, 3)	0
conv2d_4 (Conv2D)	(None, 178, 178, 32)	896
batch_normalization_4 (BatchNormalization)	(None, 178, 178, 32)	128
max_pooling2d_4 (MaxPooling2D)	(None, 89, 89, 32)	

0				
		conv2d_5 (Conv2D)	(None, 87, 87, 64)	
18,496				
		batch_normalization_5	(None, 87, 87, 64)	
256		(BatchNormalization)		
		max_pooling2d_5 (MaxPooling2D)	(None, 43, 43, 64)	
0				
		conv2d_6 (Conv2D)	(None, 41, 41, 128)	
73,856				
		batch_normalization_6	(None, 41, 41, 128)	
512		(BatchNormalization)		
		max_pooling2d_6 (MaxPooling2D)	(None, 20, 20, 128)	
0				
		conv2d_7 (Conv2D)	(None, 18, 18, 256)	
295,168				
		batch_normalization_7	(None, 18, 18, 256)	
1,024		(BatchNormalization)		
		max_pooling2d_7 (MaxPooling2D)	(None, 9, 9, 256)	
0				
		global_average_pooling2d_1	(None, 256)	
0		(GlobalAveragePooling2D)		

dense_2 (Dense)	(None, 128)	
32,896		
dropout_1 (Dropout)	(None, 128)	
0		
dense_3 (Dense)	(None, 3)	
387		

Total params: 423,619 (1.62 MB)

Trainable params: 422,659 (1.61 MB)

Non-trainable params: 960 (3.75 KB)

```
from tensorflow.keras.callbacks import EarlyStopping
```

```
early_stopping = EarlyStopping(monitor='val_accuracy',
                                patience=7,
                                mode='max')
```

```
history = model.fit(train_ds,
                    epochs=50,
                    validation_data=val_ds,
                    callbacks=[early_stopping])
```

Epoch 1/50

16/16 ————— 13s 555ms/step - accuracy: 0.5065 - loss: 1.1305 - val_accuracy: 0.4259 - val_loss: 1.0868

Epoch 2/50

16/16 ————— 7s 443ms/step - accuracy: 0.8666 - loss: 0.3589 - val_accuracy: 0.4259 - val_loss: 1.0759

Epoch 3/50

16/16 ————— 7s 443ms/step - accuracy: 0.9323 - loss: 0.1850 - val_accuracy: 0.4630 - val_loss: 1.0813

Epoch 4/50

16/16 ————— 7s 447ms/step - accuracy: 0.9626 - loss: 0.1353 - val_accuracy: 0.4630 - val_loss: 1.1073

Epoch 5/50

16/16 ————— 7s 443ms/step - accuracy: 0.9665 - loss: 0.1258 - val_accuracy: 0.5000 - val_loss: 1.1509

Epoch 6/50

16/16 ————— 7s 442ms/step - accuracy: 0.9786 - loss: 0.0842 - val_accuracy: 0.4630 - val_loss: 1.2048

Epoch 7/50

16/16 _____ 7s 445ms/step - accuracy: 0.9957 - loss: 0.0533 - val_accuracy: 0.6481 - val_loss: 1.2614
Epoch 8/50
16/16 _____ 7s 449ms/step - accuracy: 0.9832 - loss: 0.0667 - val_accuracy: 0.5370 - val_loss: 1.3185
Epoch 9/50
16/16 _____ 7s 448ms/step - accuracy: 0.9854 - loss: 0.0419 - val_accuracy: 0.5185 - val_loss: 1.3365
Epoch 10/50
16/16 _____ 7s 449ms/step - accuracy: 1.0000 - loss: 0.0371 - val_accuracy: 0.6111 - val_loss: 1.3747
Epoch 11/50
16/16 _____ 7s 445ms/step - accuracy: 0.9973 - loss: 0.0311 - val_accuracy: 0.5741 - val_loss: 1.4049
Epoch 12/50
16/16 _____ 7s 444ms/step - accuracy: 0.9941 - loss: 0.0399 - val_accuracy: 0.5926 - val_loss: 1.4633
Epoch 13/50
16/16 _____ 7s 445ms/step - accuracy: 0.9918 - loss: 0.0379 - val_accuracy: 0.5741 - val_loss: 1.4593
Epoch 14/50
16/16 _____ 7s 443ms/step - accuracy: 1.0000 - loss: 0.0221 - val_accuracy: 0.6667 - val_loss: 1.4816
Epoch 15/50
16/16 _____ 7s 444ms/step - accuracy: 1.0000 - loss: 0.0161 - val_accuracy: 0.7222 - val_loss: 1.4652
Epoch 16/50
16/16 _____ 7s 442ms/step - accuracy: 1.0000 - loss: 0.0146 - val_accuracy: 0.6852 - val_loss: 1.3921
Epoch 17/50
16/16 _____ 7s 441ms/step - accuracy: 1.0000 - loss: 0.0135 - val_accuracy: 0.7222 - val_loss: 1.3546
Epoch 18/50
16/16 _____ 7s 442ms/step - accuracy: 1.0000 - loss: 0.0134 - val_accuracy: 0.7222 - val_loss: 1.3295
Epoch 19/50
16/16 _____ 7s 444ms/step - accuracy: 1.0000 - loss: 0.0092 - val_accuracy: 0.7222 - val_loss: 1.2690
Epoch 20/50
16/16 _____ 7s 445ms/step - accuracy: 1.0000 - loss: 0.0096 - val_accuracy: 0.7222 - val_loss: 1.2061
Epoch 21/50
16/16 _____ 7s 450ms/step - accuracy: 1.0000 - loss: 0.0095 - val_accuracy: 0.7222 - val_loss: 1.1308
Epoch 22/50
16/16 _____ 7s 446ms/step - accuracy: 1.0000 - loss: 0.0082 - val_accuracy: 0.7407 - val_loss: 1.0207
Epoch 23/50
16/16 _____ 7s 440ms/step - accuracy: 1.0000 - loss:

```
0.0087 - val_accuracy: 0.7778 - val_loss: 0.9006
Epoch 24/50
16/16 _____ 7s 443ms/step - accuracy: 1.0000 - loss:
0.0079 - val_accuracy: 0.7593 - val_loss: 0.7974
Epoch 25/50
16/16 _____ 7s 448ms/step - accuracy: 1.0000 - loss:
0.0112 - val_accuracy: 0.7778 - val_loss: 0.6371
Epoch 26/50
16/16 _____ 7s 447ms/step - accuracy: 1.0000 - loss:
0.0108 - val_accuracy: 0.8148 - val_loss: 0.5343
Epoch 27/50
16/16 _____ 7s 440ms/step - accuracy: 1.0000 - loss:
0.0080 - val_accuracy: 0.8704 - val_loss: 0.4304
Epoch 28/50
16/16 _____ 7s 446ms/step - accuracy: 1.0000 - loss:
0.0056 - val_accuracy: 0.8889 - val_loss: 0.3447
Epoch 29/50
16/16 _____ 7s 440ms/step - accuracy: 1.0000 - loss:
0.0062 - val_accuracy: 0.8889 - val_loss: 0.2561
Epoch 30/50
16/16 _____ 7s 443ms/step - accuracy: 1.0000 - loss:
0.0051 - val_accuracy: 0.9259 - val_loss: 0.1902
Epoch 31/50
16/16 _____ 8s 474ms/step - accuracy: 1.0000 - loss:
0.0038 - val_accuracy: 0.9630 - val_loss: 0.1421
Epoch 32/50
16/16 _____ 8s 479ms/step - accuracy: 0.9984 - loss:
0.0053 - val_accuracy: 0.9630 - val_loss: 0.0910
Epoch 33/50
16/16 _____ 7s 466ms/step - accuracy: 1.0000 - loss:
0.0083 - val_accuracy: 0.9630 - val_loss: 0.0706
Epoch 34/50
16/16 _____ 7s 452ms/step - accuracy: 1.0000 - loss:
0.0066 - val_accuracy: 0.9815 - val_loss: 0.0638
Epoch 35/50
16/16 _____ 7s 437ms/step - accuracy: 0.9973 - loss:
0.0056 - val_accuracy: 0.9815 - val_loss: 0.0410
Epoch 36/50
16/16 _____ 7s 435ms/step - accuracy: 1.0000 - loss:
0.0105 - val_accuracy: 0.9815 - val_loss: 0.0609
Epoch 37/50
16/16 _____ 7s 450ms/step - accuracy: 0.9929 - loss:
0.0143 - val_accuracy: 1.0000 - val_loss: 0.0170
Epoch 38/50
16/16 _____ 7s 445ms/step - accuracy: 1.0000 - loss:
0.0070 - val_accuracy: 1.0000 - val_loss: 0.0141
Epoch 39/50
16/16 _____ 7s 443ms/step - accuracy: 1.0000 - loss:
0.0065 - val_accuracy: 1.0000 - val_loss: 0.0119
```

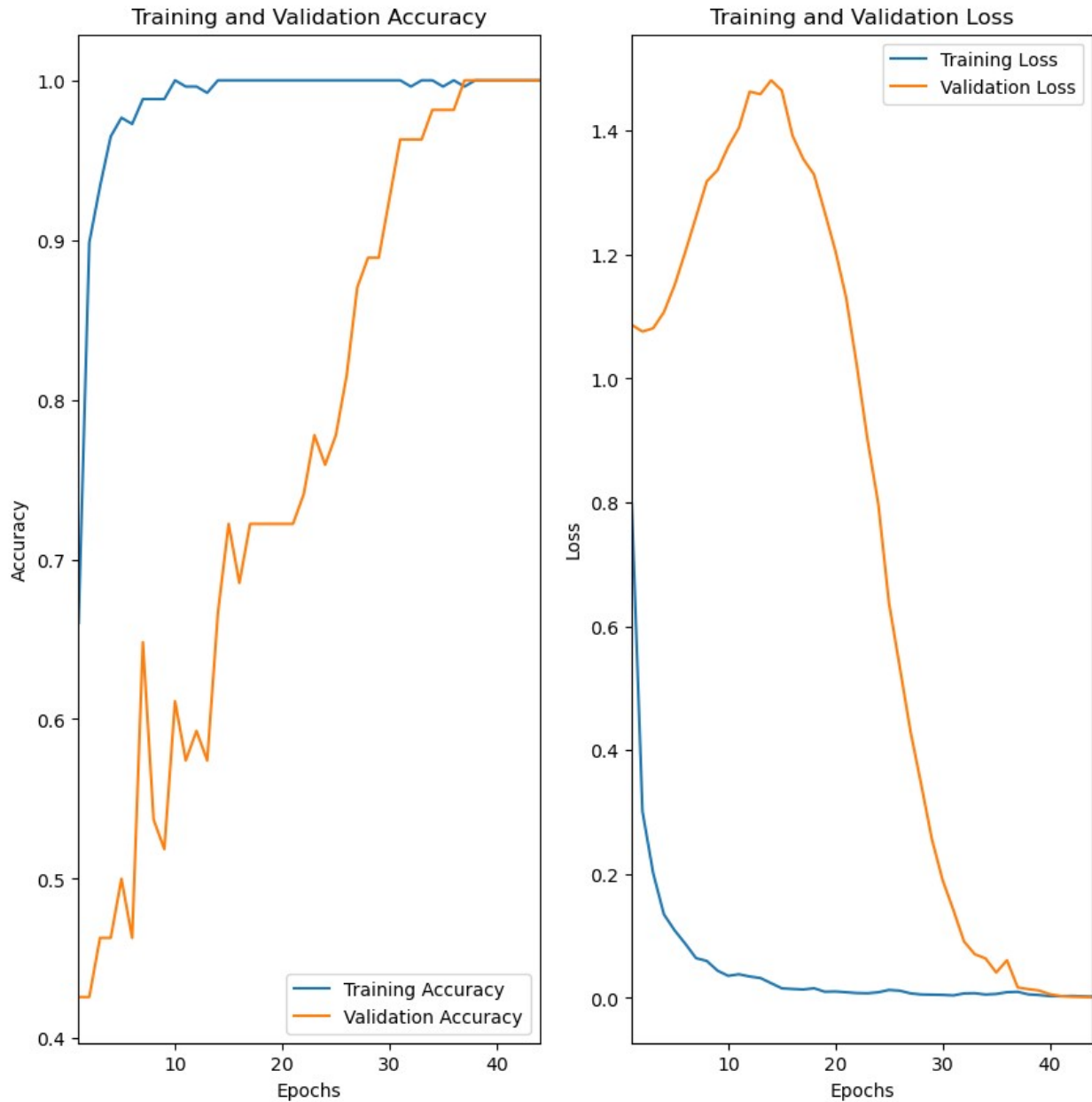
```
Epoch 40/50
16/16 _____ 7s 447ms/step - accuracy: 1.0000 - loss:
0.0023 - val_accuracy: 1.0000 - val_loss: 0.0061
Epoch 41/50
16/16 _____ 7s 438ms/step - accuracy: 1.0000 - loss:
0.0025 - val_accuracy: 1.0000 - val_loss: 0.0031
Epoch 42/50
16/16 _____ 7s 437ms/step - accuracy: 1.0000 - loss:
0.0035 - val_accuracy: 1.0000 - val_loss: 0.0019
Epoch 43/50
16/16 _____ 7s 440ms/step - accuracy: 1.0000 - loss:
0.0023 - val_accuracy: 1.0000 - val_loss: 0.0016
Epoch 44/50
16/16 _____ 7s 437ms/step - accuracy: 1.0000 - loss:
0.0017 - val_accuracy: 1.0000 - val_loss: 0.0015
```

```
epochs_range = range(1, len(history.history['loss']) + 1)

plt.figure(figsize=(10, 10))

plt.subplot(1, 2, 1)
plt.plot(epochs_range, history.history['accuracy'], label='Training
Accuracy')
plt.plot(epochs_range, history.history['val_accuracy'],
label='Validation Accuracy')
plt.legend(loc='lower right')
plt.xlim(1, len(epochs_range))
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title('Training and Validation Accuracy')

plt.subplot(1, 2, 2)
plt.plot(epochs_range, history.history['loss'], label='Training Loss')
plt.plot(epochs_range, history.history['val_loss'], label='Validation
Loss')
plt.legend(loc='upper right')
plt.xlim(1, len(epochs_range))
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Training and Validation Loss')
plt.show()
```



```
model.save('MobileNet_Bokeh.h5')
```

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')`.

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.models import load_model
```

```

from PIL import Image

model = load_model(r'C:\Users\lenovo\OneDrive\Documents\TubesPMDPM\
MobileNet_Bokeh.h5')
class_names = ['Merah', 'Thompson', 'Concord']

def classify_images(image_path, save_path='predicted_image.jpg'):
    try:
        input_image = tf.keras.utils.load_img(image_path,
target_size=(180, 180))
        input_image_array = tf.keras.utils.img_to_array(input_image)
        input_image_exp_dim = tf.expand_dims(input_image_array, 0)

        predictions = model.predict(input_image_exp_dim)
        result = tf.nn.softmax(predictions[0])
        class_idx = np.argmax(result)
        confidence = np.max(result) * 100

        print(f"Prediksi: {class_names[class_idx]}")
        print(f"Confidence: {confidence:.2f}%")

        input_image = Image.open(image_path)
        input_image.save(save_path)

        return f"Prediksi: {class_names[class_idx]} dengan confidence
{confidence:.2f}%. Gambar asli disimpan di {save_path}."
    except Exception as e:
        return f"Terjadi kesalahan: {e}"

result = classify_images(r'C:\Users\lenovo\OneDrive\Documents\
TubesPMDPM\test_data\Concord\Concord_Grape_Original_Data010.jpg',
save_path='predicted_image.jpg')
print(result)

WARNING:absl:Compiled the loaded model, but the compiled metrics have
yet to be built. `model.compile_metrics` will be empty until you train
or evaluate the model.

1/1 ————— 0s 162ms/step
Prediksi: Concord
Confidence: 57.43%
Prediksi: Concord dengan confidence 57.43%. Gambar asli disimpan di
predicted_image.jpg.

import tensorflow as tf
from tensorflow.keras.models import load_model
import seaborn as sns
import matplotlib.pyplot as plt

model = load_model(r'C:\Users\lenovo\OneDrive\Documents\TubesPMDPM\
MobileNet_Bokeh.h5')

```

```

test_data = tf.keras.preprocessing.image_dataset_from_directory(
    r'test_data',
    labels='inferred',
    label_mode='categorical',
    batch_size=32,
    image_size=(180, 180)
)

y_pred = model.predict(test_data)
y_pred_class = tf.argmax(y_pred, axis=1)

true_labels = []
for _, labels in test_data:
    true_labels.extend(tf.argmax(labels, axis=1).numpy())
true_labels = tf.convert_to_tensor(true_labels)

conf_mat = tf.math.confusion_matrix(true_labels, y_pred_class)

accuracy = tf.reduce_sum(tf.linalg.diag_part(conf_mat)) /
tf.reduce_sum(conf_mat)

precision = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=0)
recall = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=1)

f1_score = 2 * (precision * recall) / (precision + recall)

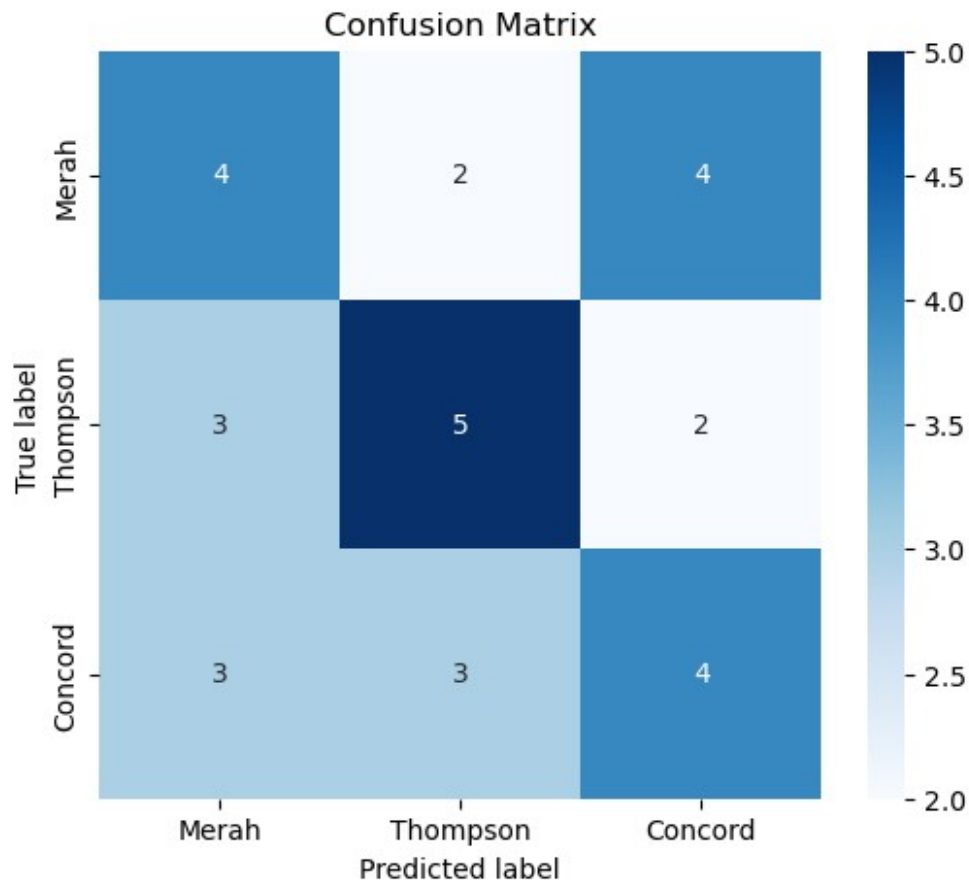
plt.figure(figsize=(6, 5))
sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
            xticklabels=["Merah", "Thompson", "Concord"],
            yticklabels=["Merah", "Thompson", "Concord"])
plt.title('Confusion Matrix')
plt.xlabel('Predicted label')
plt.ylabel('True label')
plt.show()

print("Confusion Matrix:\n", conf_mat.numpy())
print("Akurasi:", accuracy.numpy())
print("Presisi:", precision.numpy())
print("Recall:", recall.numpy())
print("F1 Score:", f1_score.numpy())

```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

Found 30 files belonging to 3 classes.
1/1 ————— 1s 676ms/step



Confusion Matrix:

```
[[4 2 4]
```

```
[3 5 2]
```

```
[3 3 4]]
```

Akurasi: 0.43333333333333335

Presisi: [0.4 0.5 0.4]

Recall: [0.4 0.5 0.4]

F1 Score: [0.4 0.5 0.4]

Panji Anugrah Agung 220711843 Bokeh Mengklasifikasi Anggur merah, anggur concord, dan anggur Thompson GoogleNet

```
import tensorflow as tf
import cv2
import numpy as np
from matplotlib import pyplot as plt
#load data
data_dir = r"C:\Users\bravo\Downloads\Documents\pembelajaranMesin\
train_data"
#Randomize data yang telah di load sekaligus resize menjadi 180 x 180
data = tf.keras.utils.image_dataset_from_directory(data_dir, seed=
123, image_size=(180,180), batch_size=16)
print(data.class_names)

class_names = data.class_names

img_size = 180
batch = 10
validation_split = 0.1
dataset = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
)

total_count = len(dataset)
val_count = int(total_count * validation_split)
train_count = total_count - val_count

print("Total Images:", total_count)
print("Train Images:", train_count)
print("Validation Images:", val_count)

train_ds = dataset.take(train_count)
val_ds = dataset.skip(train_count)

Found 322 files belonging to 3 classes.
['Concord', 'Merah', 'Thompson']
Found 322 files belonging to 3 classes.
Total Images: 33
Train Images: 30
Validation Images: 3

import matplotlib.pyplot as plt

i = 0
plt.figure(figsize=(10,10))
```

```
#tampilkan untuk memastikan data sudah di Load
for images, labels in train_ds.take(1):
    for i in range(9):
        plt.subplot(3,3, i+1)
        plt.imshow(images[i].numpy().astype('uint8'))
        plt.title(class_names[labels[i]])
        plt.axis('off')
```

Concord



Thompson



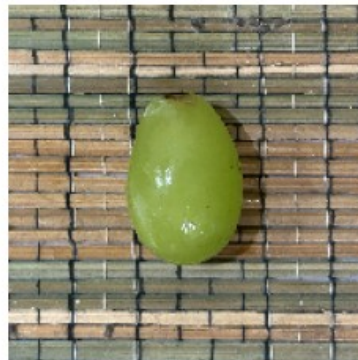
Concord



Merah



Thompson



Concord



Thompson



Concord



Merah



```
for images, labels in train_ds.take(1):
    images_array = np.array(images)
    print(images_array.shape)
```

```
(10, 180, 180, 3)

import matplotlib.pyplot as plt

i = 0
plt.figure(figsize=(10,10))

#tampilkan untuk memastikan data sudah di load
for images, labels in train_ds.take(1):
    for i in range(9):
        plt.subplot(3,3, i+1)
        plt.imshow(images[i].numpy().astype('uint8'))
        plt.axis('off')
```



```
for images, labels in train_ds.take(1):  
    images_array = np.array(images)  
    print(images_array.shape)
```

#loop untuk mengecek atribut gambar(jumlah, tinggi, lebar, dan channel(RGB))

```
(10, 180, 180, 3)
```

```
from tensorflow.keras import layers  
from tensorflow.keras.models import Sequential, load_model
```

```
Tuner = tf.data.AUTOTUNE
```



```
train_ds = train_ds.cache().shuffle(1000).prefetch(buffer_size =
Tuner)
val_ds = val_ds.cache().shuffle(1000).prefetch(buffer_size = Tuner)
```

#Augmentasi data dengan menggunakan Sequential

```
data_augmentation = Sequential([
    layers.RandomFlip("horizontal", input_shape =
(img_size,img_size,3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1),
    layers.RandomBrightness(0.1)
])
```

```
i = 0
```

```
plt.figure(figsize=(10,10))
```

#Lihat data setelah di augmentasi

```
for images, labels in train_ds.take(69):
    for i in range(9):
        images = data_augmentation(images)
        plt.subplot(3,3, i+1)
        plt.imshow(images[0].numpy().astype('uint8'))
        plt.axis('off')
```

C:\Users\bravo\AppData\Roaming\Python\Python312\site-packages\keras\src\layers\preprocessing\tf_data_layer.py:19: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(**kwargs)
```




```
import tensorflow as tf
import keras

import keras._tf_keras.keras.backend as K
from keras._tf_keras.keras.models import Model
from keras._tf_keras.keras.layers import Input, Dense, Conv2D
from keras._tf_keras.keras.layers import Flatten, MaxPool2D, AvgPool2D
from keras._tf_keras.keras.layers import Concatenate, Dropout

from keras._tf_keras.keras.models import load_model

#membuat model from scratch
```

```

def googlenet(input_shape, n_classes):

    def inception_block(x, f):
        t1 = Conv2D(f[0], 1, activation='relu')(x)

        t2 = Conv2D(f[1], 1, activation='relu')(x)
        t2 = Conv2D(f[2], 3, padding='same', activation='relu')(t2)

        t3 = Conv2D(f[3], 1, activation='relu')(x)
        t3 = Conv2D(f[4], 5, padding='same', activation='relu')(t3)

        t4 = MaxPool2D(3, 1, padding='same')(x)
        t4 = Conv2D(f[5], 1, activation='relu')(t4)

        output = Concatenate()([t1, t2, t3, t4])
        return output

    input = Input(input_shape)

    x = Conv2D(64, 7, strides=2, padding='same', activation='relu')(
input)
    x = MaxPool2D(3, strides=2, padding='same')(x)

    x = Conv2D(64, 1, activation='relu')(x)
    x = Conv2D(192, 3, padding='same', activation='relu')(x)
    x = MaxPool2D(3, strides=2)(x)

    x = inception_block(x, [64, 96, 128, 16, 32, 32])
    x = inception_block(x, [128, 128, 192, 32, 96, 64])
    x = MaxPool2D(3, strides=2, padding='same')(x)

    x = inception_block(x, [192, 96, 208, 16, 48, 64])
    x = inception_block(x, [160, 112, 224, 24, 64, 64])
    x = inception_block(x, [128, 128, 256, 24, 64, 64])
    x = inception_block(x, [112, 144, 288, 32, 64, 64])
    x = inception_block(x, [256, 160, 320, 32, 128, 128])
    x = MaxPool2D(3, strides=2, padding='same')(x)

    x = inception_block(x, [256, 160, 320, 32, 128, 128])
    x = inception_block(x, [384, 192, 384, 48, 128, 128])

    x = AvgPool2D(3, strides=1)(x)
    x = Dropout(0.4)(x)

    x = Flatten()(x)
    output = Dense(n_classes, activation='softmax')(x)

    model = Model(input, output)
    return model

```

```
#Pastikan input shae dan jumlah kelas sesuai
input_shape = 180, 180, 3
n_classes = 3

#Clear Cache Keras menggunakan clear session
K.clear_session()
#buat model dengan
model = googlenet(input_shape, n_classes)
model.summary()
###Terdapat code yang hilang disini! lihat modul untuk menemukanya

Model: "functional"
```

Layer (type)	Output Shape	Param #	Connected to
input_layer (InputLayer)	(None, 180, 180, 3)	0	-
conv2d (Conv2D) input_layer[0][0]	(None, 90, 90, 64)	9,472	
max_pooling2d (MaxPooling2D)	(None, 45, 45, 64)	0	conv2d[0][0]
conv2d_1 (Conv2D) max_pooling2d[0]...	(None, 45, 45, 64)	4,160	
conv2d_2 (Conv2D) [0]	(None, 45, 45, 192)	110,784	conv2d_1[0]
max_pooling2d_1	(None, 22, 22, ...)	0	conv2d_2[0]

[0]	(MaxPooling2D)	192)		
conv2d_4 (Conv2D)	(None, 22, 22,	18,528		
max_pooling2d_1[...]	96)			
conv2d_6 (Conv2D)	(None, 22, 22,	3,088		
max_pooling2d_1[...]	16)			
max_pooling2d_2	(None, 22, 22,	0		
max_pooling2d_1[...]	(MaxPooling2D)	192)		
conv2d_3 (Conv2D)	(None, 22, 22,	12,352		
max_pooling2d_1[...]	64)			
conv2d_5 (Conv2D)	(None, 22, 22,	110,720	conv2d_4[0]	
[0]	128)			
conv2d_7 (Conv2D)	(None, 22, 22,	12,832	conv2d_6[0]	
[0]	32)			
conv2d_8 (Conv2D)	(None, 22, 22,	6,176		
max_pooling2d_2[...]	32)			
concatenate	(None, 22, 22,	0	conv2d_3[0]	
[0],				

(Concatenate)	256)		conv2d_5[0]
[0],			
[0],			conv2d_7[0]
[0]			conv2d_8[0]
conv2d_10 (Conv2D)	(None, 22, 22,	32,896	
concatenate[0][0]	128)		
conv2d_12 (Conv2D)	(None, 22, 22,	8,224	
concatenate[0][0]	32)		
max_pooling2d_3	(None, 22, 22,	0	
concatenate[0][0]	(MaxPooling2D)	256)	
conv2d_9 (Conv2D)	(None, 22, 22,	32,896	
concatenate[0][0]	128)		
conv2d_11 (Conv2D)	(None, 22, 22,	221,376	conv2d_10[0]
[0]	192)		
conv2d_13 (Conv2D)	(None, 22, 22,	76,896	conv2d_12[0]
[0]	96)		
conv2d_14 (Conv2D)	(None, 22, 22,	16,448	
max_pooling2d_3[...]	64)		

concatenate_1 [0], (Concatenate)	(None, 22, 22, 480)	0	conv2d_9[0]
conv2d_11[0]			conv2d_13[0]
conv2d_14[0]			conv2d_14[0]
max_pooling2d_4 concatenate_1[0]... (MaxPooling2D)	(None, 11, 11, 480)	0	
conv2d_16 (Conv2D) max_pooling2d_4[...]	(None, 11, 11, 96)	46,176	
conv2d_18 (Conv2D) max_pooling2d_4[...]	(None, 11, 11, 16)	7,696	
max_pooling2d_5 max_pooling2d_4[...] (MaxPooling2D)	(None, 11, 11, 480)	0	
conv2d_15 (Conv2D) max_pooling2d_4[...]	(None, 11, 11, 192)	92,352	
conv2d_17 (Conv2D) [0]	(None, 11, 11, 208)	179,920	conv2d_16[0]
conv2d_19 (Conv2D) [0]	(None, 11, 11, 19,248)		conv2d_18[0]

	48)		
conv2d_20 (Conv2D)	(None, 11, 11,	30,784	
max_pooling2d_5[...]	64)		
concatenate_2	(None, 11, 11,	0	conv2d_15[0]
[0],			
(Concatenate)	512)		conv2d_17[0]
[0],			
[0],			conv2d_19[0]
[0]			conv2d_20[0]
conv2d_22 (Conv2D)	(None, 11, 11,	57,456	
concatenate_2[0]...	112)		
conv2d_24 (Conv2D)	(None, 11, 11,	12,312	
concatenate_2[0]...	24)		
max_pooling2d_6	(None, 11, 11,	0	
concatenate_2[0]...			
(MaxPooling2D)	512)		
conv2d_21 (Conv2D)	(None, 11, 11,	82,080	
concatenate_2[0]...	160)		
conv2d_23 (Conv2D)	(None, 11, 11,	226,016	conv2d_22[0]
[0]	224)		

conv2d_25 (Conv2D)	(None, 11, 11, 64)	38,464	conv2d_24[0]
conv2d_26 (Conv2D)	(None, 11, 11, 64)	32,832	
concatenate_3	(None, 11, 11, 512)	0	conv2d_21[0]
			conv2d_23[0]
			conv2d_25[0]
			conv2d_26[0]
conv2d_28 (Conv2D)	(None, 11, 11, 128)	65,664	
conv2d_30 (Conv2D)	(None, 11, 11, 24)	12,312	
max_pooling2d_7	(None, 11, 11, 512)	0	
conv2d_27 (Conv2D)	(None, 11, 11, 128)	65,664	
conv2d_29 (Conv2D)	(None, 11, 11, 295,168)		conv2d_28[0]

	256)		
conv2d_31 (Conv2D)	(None, 11, 11,	38,464	conv2d_30[0]
	64)		
conv2d_32 (Conv2D)	(None, 11, 11,	32,832	
max_pooling2d_7[...]	64)		
concatenate_4	(None, 11, 11,	0	conv2d_27[0]
[0],	(Concatenate)	512)	conv2d_29[0]
[0],			conv2d_31[0]
[0],			conv2d_32[0]
[0]			
conv2d_34 (Conv2D)	(None, 11, 11,	73,872	
concatenate_4[0]...	144)		
conv2d_36 (Conv2D)	(None, 11, 11,	16,416	
concatenate_4[0]...	32)		
max_pooling2d_8	(None, 11, 11,	0	
concatenate_4[0]...	(MaxPooling2D)	512)	
conv2d_33 (Conv2D)	(None, 11, 11,	57,456	
concatenate_4[0]...	112)		

conv2d_35 (Conv2D)	(None, 11, 11, 288)	373,536	conv2d_34[0]
conv2d_37 (Conv2D)	(None, 11, 11, 64)	51,264	conv2d_36[0]
conv2d_38 (Conv2D) max_pooling2d_8[0]	(None, 11, 11, 64)	32,832	
concatenate_5 [0], (Concatenate) [0], [0], [0]	(None, 11, 11, 528)	0	conv2d_33[0] conv2d_35[0] conv2d_37[0] conv2d_38[0]
conv2d_40 (Conv2D) concatenate_5[0]	(None, 11, 11, 160)	84,640	
conv2d_42 (Conv2D) concatenate_5[0]	(None, 11, 11, 32)	16,928	
max_pooling2d_9 concatenate_5[0]	(None, 11, 11, 528)	0	
conv2d_39 (Conv2D) concatenate_5[0]	(None, 11, 11, 135,424)		

	256)		
conv2d_41 (Conv2D)	(None, 11, 11,	461,120	conv2d_40[0]
	320)		
conv2d_43 (Conv2D)	(None, 11, 11,	102,528	conv2d_42[0]
	128)		
conv2d_44 (Conv2D)	(None, 11, 11,	67,712	
max_pooling2d_9[...]	128)		
concatenate_6	(None, 11, 11,	0	conv2d_39[0]
(Concatenate)	832)		conv2d_41[0]
			conv2d_43[0]
			conv2d_44[0]
max_pooling2d_10	(None, 6, 6, 832)	0	
concatenate_6[0]...			
(MaxPooling2D)			
conv2d_46 (Conv2D)	(None, 6, 6, 160)	133,280	
max_pooling2d_10...			
conv2d_48 (Conv2D)	(None, 6, 6, 32)	26,656	
max_pooling2d_10...			
max_pooling2d_11	(None, 6, 6, 832)	0	
max_pooling2d_10...			
(MaxPooling2D)			

conv2d_45 (Conv2D)	(None, 6, 6, 256)	213,248	
max_pooling2d_10...			
conv2d_47 (Conv2D)	(None, 6, 6, 320)	461,120	conv2d_46[0]
conv2d_49 (Conv2D)	(None, 6, 6, 128)	102,528	conv2d_48[0]
conv2d_50 (Conv2D)	(None, 6, 6, 128)	106,624	
max_pooling2d_11...			
concatenate_7	(None, 6, 6, 832)	0	conv2d_45[0],
(Concatenate)			conv2d_47[0]
			conv2d_49[0]
			conv2d_50[0]
conv2d_52 (Conv2D)	(None, 6, 6, 192)	159,936	
concatenate_7[0]...			
conv2d_54 (Conv2D)	(None, 6, 6, 48)	39,984	
concatenate_7[0]...			
max_pooling2d_12	(None, 6, 6, 832)	0	
concatenate_7[0]...			
(MaxPooling2D)			
conv2d_51 (Conv2D)	(None, 6, 6, 384)	319,872	
concatenate_7[0]...			
conv2d_53 (Conv2D)	(None, 6, 6, 384)	663,936	conv2d_52[0]

conv2d_55 (Conv2D)	(None, 6, 6, 128)	153,728	conv2d_54[0]
conv2d_56 (Conv2D)	(None, 6, 6, 128)	106,624	
concatenate_8	(None, 6, 6, 1024)	0	conv2d_51[0], conv2d_53[0], conv2d_55[0], conv2d_56[0]
average_pooling2d	(None, 4, 4, 1024)	0	
dropout (Dropout)	(None, 4, 4, 1024)	0	
flatten (Flatten)	(None, 16384)	0	dropout[0][0]
dense (Dense)	(None, 3)	49,155	flatten[0][0]

Total params: 6,022,707 (22.97 MB)

Trainable params: 6,022,707 (22.97 MB)

Non-trainable params: 0 (0.00 B)

```
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.optimizers import Adam
#Coimpile dengan optimizer adam
```

```

###Terdapat code yang hilang disini! lihat modul untuk menemukanya
model.compile(
    optimizer=Adam(),
    loss='sparse_categorical_crossentropy',
    metrics=['accuracy']
)

```

```

#buat early stopping
early_stopping = EarlyStopping(monitor='val_accuracy',
                                patience=5,
                                mode='max')

```

```

#fit validation data ke dalam model
history= model.fit(train_ds,
                    epochs=30,
                    validation_data=val_ds,
                    callbacks=[early_stopping])

```

```

Epoch 1/30
30/30 _____ 26s 350ms/step - accuracy: 0.3696 - loss:
5.8330 - val_accuracy: 0.5000 - val_loss: 1.0771
Epoch 2/30
30/30 _____ 9s 288ms/step - accuracy: 0.3932 - loss:
1.1052 - val_accuracy: 0.2727 - val_loss: 1.1007
Epoch 3/30
30/30 _____ 9s 286ms/step - accuracy: 0.3212 - loss:
1.1001 - val_accuracy: 0.2727 - val_loss: 1.0987
Epoch 4/30
30/30 _____ 9s 284ms/step - accuracy: 0.3616 - loss:
1.0981 - val_accuracy: 0.3636 - val_loss: 1.1015
Epoch 5/30
30/30 _____ 9s 285ms/step - accuracy: 0.3548 - loss:
1.0930 - val_accuracy: 0.3182 - val_loss: 1.0917
Epoch 6/30
30/30 _____ 9s 286ms/step - accuracy: 0.3872 - loss:
1.0938 - val_accuracy: 0.2727 - val_loss: 1.1054

```

```

#buat plot dengan menggunakan history supaya jumlahnya sesuai epoch yang dilakukan

```

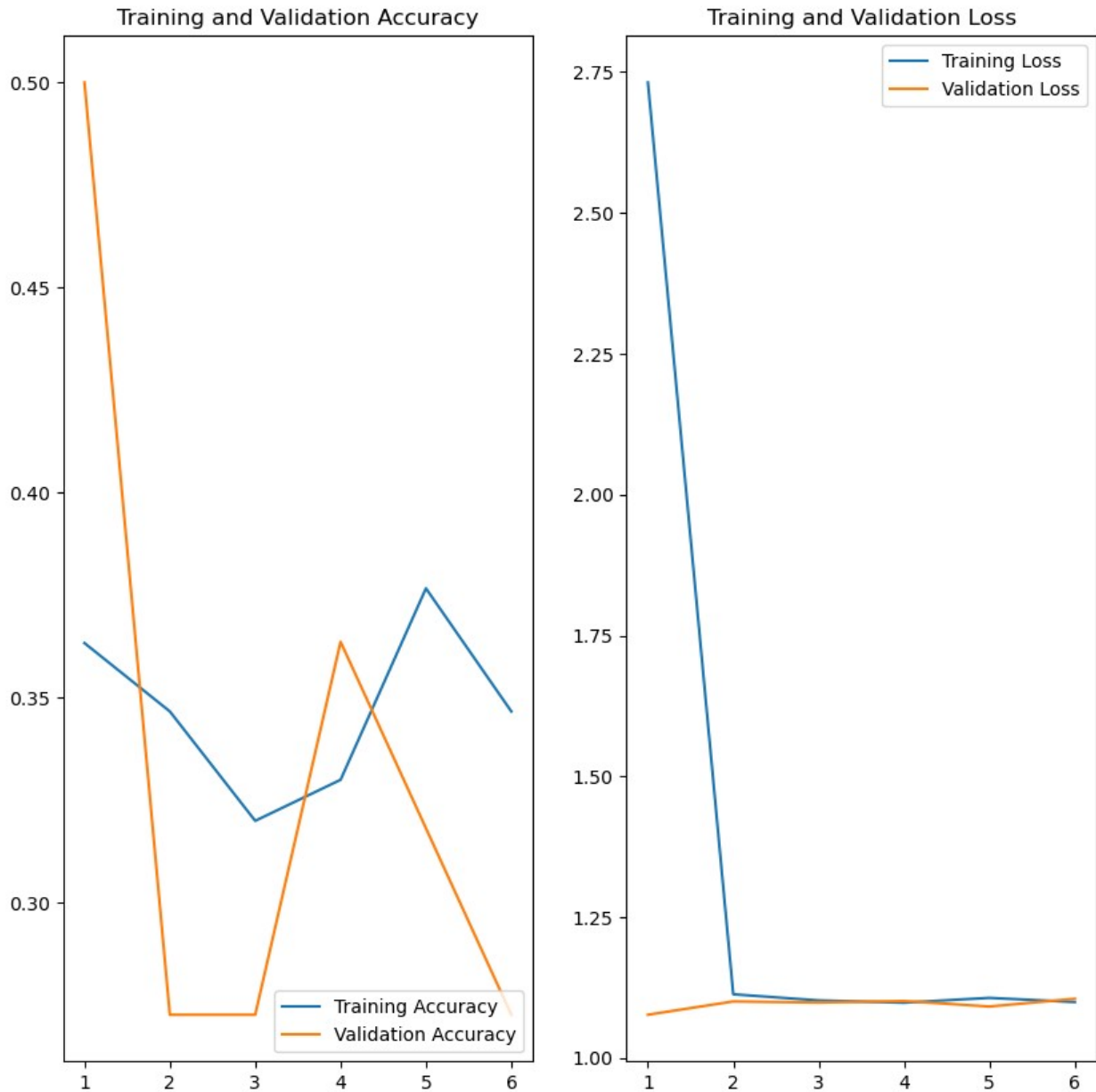
```

ephocs_range = range(1, len(history.history['loss']) + 1)
plt.figure(figsize=(10, 10))
plt.subplot(1, 2, 1)
plt.plot(ephocs_range, history.history['accuracy'], label='Training Accuracy')
plt.plot(ephocs_range, history.history['val_accuracy'],
label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')

plt.subplot(1, 2, 2)
plt.plot(ephocs_range, history.history['loss'], label='Training Loss')

```

```
plt.plot(epochs_range, history.history['val_loss'], label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()
```



```
model.save('gugelnet.h5')
```

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')`.
```

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.models import load_model
from PIL import Image

# Load the trained model
model = load_model(r'C:\Users\bravo\Downloads\Documents\
pembelajaranMesin\gugelnet.h5') # Ganti dengan path model Anda
class_names = ['Concord', 'Merah', 'Thompson']

# Function to classify images and save the original image
def classify_images(image_path, save_path='predicted_image.jpg'):
    try:
        # Load and preprocess the image
        input_image = tf.keras.utils.load_img(image_path,
target_size=(180, 180))
        input_image_array = tf.keras.utils.img_to_array(input_image)
        input_image_exp_dim = tf.expand_dims(input_image_array, 0) #
Add batch dimension

        # Predict
        predictions = model.predict(input_image_exp_dim)
        result = tf.nn.softmax(predictions[0])
        class_idx = np.argmax(result)
        confidence = np.max(result) * 100

        # Display prediction and confidence in notebook
        print(f"Prediksi: {class_names[class_idx]}")
        print(f"Confidence: {confidence:.2f}%")

        # Save the original image (without text)
        input_image = Image.open(image_path)
        input_image.save(save_path)

        return f"Prediksi: {class_names[class_idx]} dengan confidence
{confidence:.2f}%. Gambar asli disimpan di {save_path}."
    except Exception as e:
        return f"Terjadi kesalahan: {e}"

# Contoh penggunaan fungsi
###Terdapat code yang hilang disini! lihat modul untuk menemukanya
result = classify_images(r'C:\Users\bravo\Downloads\Documents\
pembelajaranMesin\test_data\Merah\Merah_Grape_Original_Data02.JPG',
save_path='mentah2.jpg')
print(result)
```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

WARNING:tensorflow:6 out of the last 6 calls to <function TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at 0x0000024E79D4D3A0> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3) passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2), @tf.function has reduce_retracing=True option that can avoid unnecessary retracing. For (3), please refer to https://www.tensorflow.org/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf/function for more details.

WARNING:tensorflow:6 out of the last 6 calls to <function TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at 0x0000024E79D4D3A0> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3) passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2), @tf.function has reduce_retracing=True option that can avoid unnecessary retracing. For (3), please refer to https://www.tensorflow.org/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf/function for more details.

1/1 ————— 1s 621ms/step

Prediksi: Merah

Confidence: 34.40%

Prediksi: Merah dengan confidence 34.40%. Gambar asli disimpan di mentah2.jpg.

```
import tensorflow as tf
from tensorflow.keras.models import load_model
import seaborn as sns
import matplotlib.pyplot as plt

# Muat data test yang sebenarnya
test_data = tf.keras.preprocessing.image_dataset_from_directory(
    r'test_data',
    labels='inferred',
    label_mode='categorical', # Menghasilkan label dalam bentuk one-
hot encoding
    batch_size=32,
    image_size=(180, 180)
)
```

```

# Prediksi model
y_pred = model.predict(test_data)
y_pred_class = tf.argmax(y_pred, axis=1) # Konversi ke kelas prediksi

# Ekstrak label sebenarnya dari test_data dan konversi ke bentuk indeks kelas
true_labels = []
for _, labels in test_data:
    true_labels.extend(tf.argmax(labels, axis=1).numpy()) # Konversi one-hot ke indeks kelas
true_labels = tf.convert_to_tensor(true_labels)

# Membuat matriks kebingungan
###Terdapat code yang hilang disini! lihat modul untuk menemukanya
conf_mat = tf.math.confusion_matrix(true_labels, y_pred_class)

# Menghitung akurasi
accuracy = tf.reduce_sum(tf.linalg.diag_part(conf_mat)) / tf.reduce_sum(conf_mat)

# Menghitung presisi dan recall
precision = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat, axis=0)
recall = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat, axis=1)

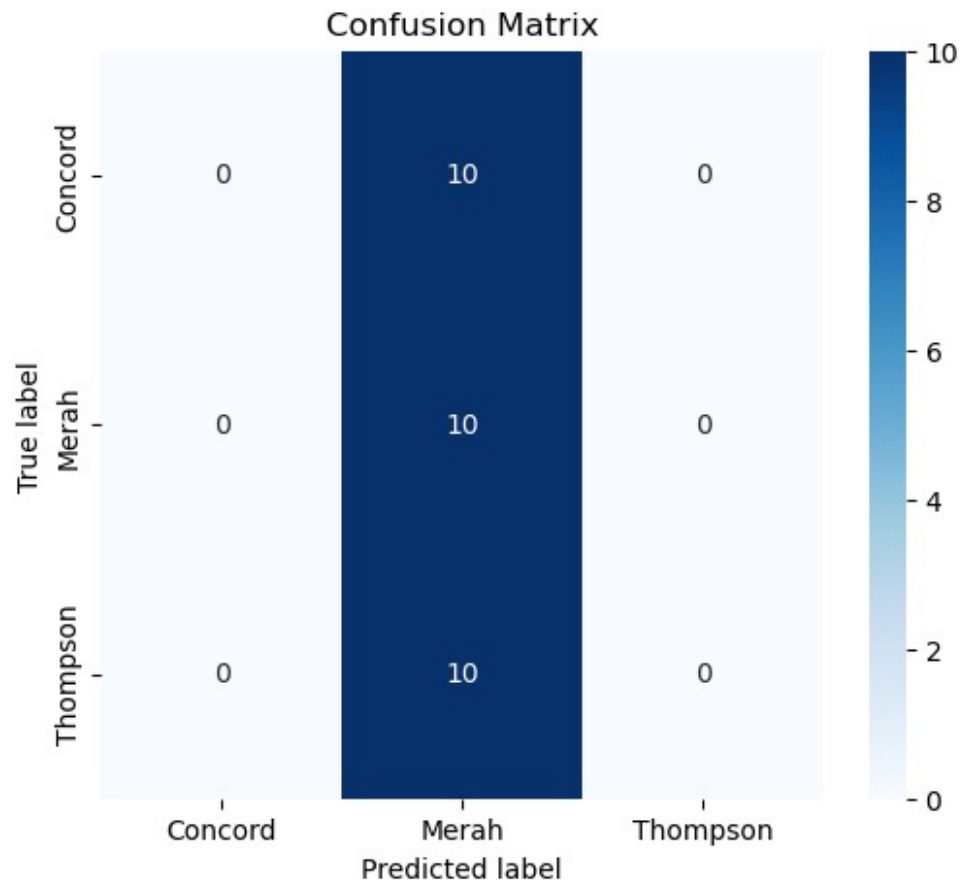
# Menghitung F1 Score
f1_score = 2 * (precision * recall) / (precision + recall)

# Visualisasi Confusion Matrix
plt.figure(figsize=(6, 5))
sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
            xticklabels=["Concord", "Merah", "Thompson"],
            yticklabels=["Concord", "Merah", "Thompson"])
plt.title('Confusion Matrix')
plt.xlabel('Predicted label')
plt.ylabel('True label')
plt.show()

# Menampilkan hasil
###Terdapat code yang hilang disini! lihat modul untuk menemukanya
print("Confusion Matrix:\n", conf_mat.numpy())
print("Akurasi:", accuracy.numpy())
print("Presisi:", precision.numpy())
print("Recall:", recall.numpy())
print("F1 Score:", f1_score.numpy())

Found 30 files belonging to 3 classes.
1/1 _____ 1s 1s/step

```



```
Confusion Matrix:
[[ 0 10  0]
 [ 0 10  0]
 [ 0 10  0]]
Akurasi: 0.3333333333333333
Presisi: [      nan 0.33333333      nan]
Recall: [0.  1.  0.]
F1 Score: [nan 0.5 nan]
```

```
import tensorflow as tf
import numpy as np
from matplotlib import pyplot as plt

data_dir = r"C:\Users\smerf\Downloads\Tugas6_B_11881\train_data"

data = tf.keras.utils.image_dataset_from_directory(data_dir, seed =
123, image_size=(180, 180), batch_size=16)
print(data.class_names)

class_names = data.class_names

img_size = 180
batch = 32
validation_split = 0.1
dataset = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
)
total_count = len(dataset)
val_count = int(total_count * validation_split)
train_count = total_count - val_count

print("Total Images:", total_count)
print("Train Images:", train_count)
print("Validation Images:", val_count)

train_ds = dataset.take(train_count)
val_ds = dataset.skip(train_count)

Found 300 files belonging to 3 classes.
['Concord', 'Merah', 'Thompson']
Found 300 files belonging to 3 classes.
Total Images: 10
Train Images: 9
Validation Images: 1

import matplotlib.pyplot as plt

i = 0
plt.figure(figsize=(10,10))

for images, labels in train_ds.take(1):
    for i in range(9):
        plt.subplot(3,3, i+1)
```



```
plt.imshow(images[i].numpy().astype('uint8'))  
plt.axis('off')
```



```
for images, labels in train_ds.take(1):  
    images_array = np.array(images)  
    print(images_array.shape)  
  
(32, 180, 180, 3)  
  
from tensorflow.keras import layers  
from tensorflow.keras.models import Sequential, load_model
```

```
Tuner = tf.data.AUTOTUNE
train_ds = train_ds.cache().shuffle(1000).prefetch(buffer_size =
Tuner)
val_ds = val_ds.cache().shuffle(1000).prefetch(buffer_size = Tuner)
```

```
data_augmentation = Sequential([
    layers.RandomFlip('horizontal', input_shape =
(img_size,img_size,3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1)
])
```

```
i = 0
plt.figure(figsize=(10,10))
#Lihat data setelah di augmentasi
for images, labels in train_ds.take(69):
    for i in range(9):
        images = data_augmentation(images)
        plt.subplot(3,3, i+1)
        plt.imshow(images[0].numpy().astype('uint8'))
        plt.axis('off')
```

```
C:\Users\smerf\AppData\Local\Packages\
PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\LocalCache\local-
packages\Python311\site-packages\keras\src\layers\preprocessing\
tf_data_layer.py:19: UserWarning: Do not pass an
`input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in
the model instead.
    super().__init__(**kwargs)
```



```
import tensorflow as tf
from tensorflow.keras import layers, models
import keras._tf_keras.keras.backend as K

def vgg16(input_shape, n_classes):
    model = models.Sequential()

    model.add(layers.Conv2D(64, (3, 3), activation='relu',
padding='same', input_shape=input_shape))
    model.add(layers.Conv2D(64, (3, 3), activation='relu',
padding='same'))
    model.add(layers.MaxPooling2D((2, 2), strides=(2, 2)))
```

```

        model.add(layers.Conv2D(128, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(128, (3, 3), activation='relu',
padding='same'))
        model.add(layers.MaxPooling2D((2, 2), strides=(2, 2)))

        model.add(layers.Conv2D(256, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(256, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(256, (3, 3), activation='relu',
padding='same'))
        model.add(layers.MaxPooling2D((2, 2), strides=(2, 2)))

        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.MaxPooling2D((2, 2), strides=(2, 2)))

        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.Conv2D(512, (3, 3), activation='relu',
padding='same'))
        model.add(layers.MaxPooling2D((2, 2), strides=(2, 2)))

        model.add(layers.Flatten())
        model.add(layers.Dense(4096, activation='relu'))
        model.add(layers.Dropout(0.5))
        model.add(layers.Dense(4096, activation='relu'))
        model.add(layers.Dropout(0.5))
        model.add(layers.Dense(n_classes, activation='softmax'))

    return model

input_shape = (180, 180, 3)
n_classes = 2

K.clear_session()

model = vgg16(input_shape, n_classes)
model.summary()

```

WARNING:tensorflow:From C:\Users\smerf\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\LocalCache\local-

```
packages\Python311\site-packages\keras\src\backend\common\
global_state.py:82: The name tf.reset_default_graph is deprecated.
Please use tf.compat.v1.reset_default_graph instead.
```

```
C:\Users\smerf\AppData\Local\Packages\
PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\LocalCache\local-
packages\Python311\site-packages\keras\src\layers\convolutional\
base_conv.py:107: UserWarning: Do not pass an
`input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in
the model instead.
```

```
super().__init__(activity_regularizer=activity_regularizer,
**kwargs)
```

Model: "sequential"

Layer (type) Param #	Output Shape	
conv2d (Conv2D) 1,792	(None, 180, 180, 64)	
conv2d_1 (Conv2D) 36,928	(None, 180, 180, 64)	
max_pooling2d (MaxPooling2D) 0	(None, 90, 90, 64)	
conv2d_2 (Conv2D) 73,856	(None, 90, 90, 128)	
conv2d_3 (Conv2D) 147,584	(None, 90, 90, 128)	
max_pooling2d_1 (MaxPooling2D) 0	(None, 45, 45, 128)	
conv2d_4 (Conv2D) 295,168	(None, 45, 45, 256)	

conv2d_5 (Conv2D)	(None, 45, 45, 256)	
590,080		
conv2d_6 (Conv2D)	(None, 45, 45, 256)	
590,080		
max_pooling2d_2 (MaxPooling2D)	(None, 22, 22, 256)	
0		
conv2d_7 (Conv2D)	(None, 22, 22, 512)	
1,180,160		
conv2d_8 (Conv2D)	(None, 22, 22, 512)	
2,359,808		
conv2d_9 (Conv2D)	(None, 22, 22, 512)	
2,359,808		
max_pooling2d_3 (MaxPooling2D)	(None, 11, 11, 512)	
0		
conv2d_10 (Conv2D)	(None, 11, 11, 512)	
2,359,808		
conv2d_11 (Conv2D)	(None, 11, 11, 512)	
2,359,808		
conv2d_12 (Conv2D)	(None, 11, 11, 512)	
2,359,808		
max_pooling2d_4 (MaxPooling2D)	(None, 5, 5, 512)	
0		
flatten (Flatten)	(None, 12800)	
0		

dense (Dense)	(None, 4096)	
52,432,896		
dropout (Dropout)	(None, 4096)	
0		
dense_1 (Dense)	(None, 4096)	
16,781,312		
dropout_1 (Dropout)	(None, 4096)	
0		
dense_2 (Dense)	(None, 2)	
8,194		

Total params: 83,937,090 (320.19 MB)

Trainable params: 83,937,090 (320.19 MB)

Non-trainable params: 0 (0.00 B)

```
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.optimizers import Adam
```

```
n_classes = 3
```

```
model = vgg16(input_shape, n_classes)
```

```
model.compile(
    optimizer=Adam(),
    loss='sparse_categorical_crossentropy',
)
```

```
early_stopping = EarlyStopping(monitor='val_accuracy',
                                patience=5,
                                mode='max')
```

```
history = model.fit(train_ds,
                    epochs=30,
                    validation_data=val_ds,
                    callbacks=[early_stopping])
```

Epoch 1/30

9/9 _____ 115s 12s/step - accuracy: 0.3861 - loss: 197.9797 - val_accuracy: 0.5833 - val_loss: 1.5339

Epoch 2/30
9/9 _____ 105s 12s/step - accuracy: 0.3919 - loss: 1.3644 - val_accuracy: 0.1667 - val_loss: 1.1038

Epoch 3/30
9/9 _____ 109s 12s/step - accuracy: 0.2777 - loss: 1.1617 - val_accuracy: 0.5833 - val_loss: 1.0776

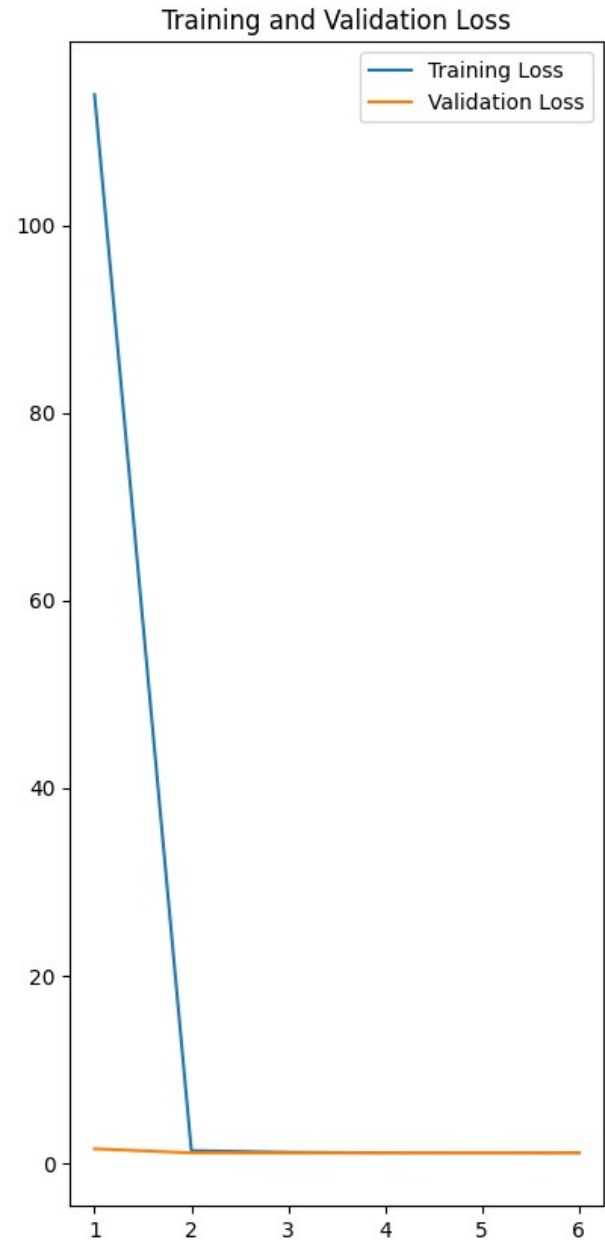
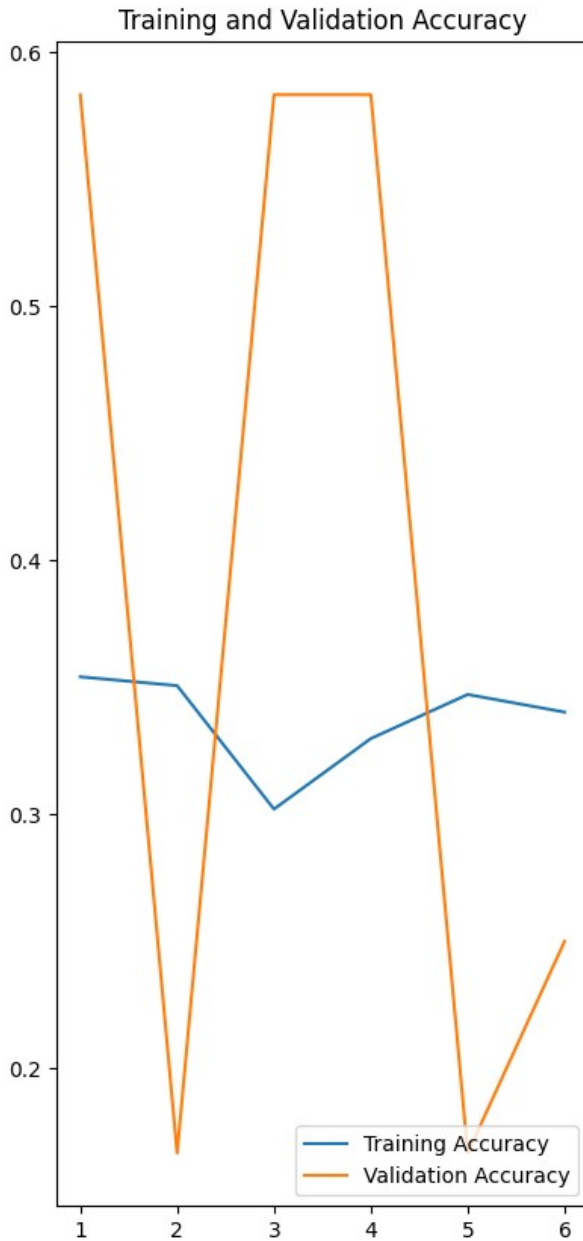
Epoch 4/30
9/9 _____ 106s 12s/step - accuracy: 0.3198 - loss: 1.1015 - val_accuracy: 0.5833 - val_loss: 1.0951

Epoch 5/30
9/9 _____ 100s 11s/step - accuracy: 0.3596 - loss: 1.0998 - val_accuracy: 0.1667 - val_loss: 1.1126

Epoch 6/30
9/9 _____ 107s 12s/step - accuracy: 0.3625 - loss: 1.0999 - val_accuracy: 0.2500 - val_loss: 1.0935

```
ephocs_range = range(1, len(history.history['loss']) + 1)
plt.figure(figsize=(10, 10))
plt.subplot(1, 2, 1)
plt.plot(ephocs_range, history.history['accuracy'], label='Training Accuracy')
plt.plot(ephocs_range, history.history['val_accuracy'], label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')

plt.subplot(1, 2, 2)
plt.plot(ephocs_range, history.history['loss'], label='Training Loss')
plt.plot(ephocs_range, history.history['val_loss'], label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()
```

```
model.save('vgg-16.h5')
```

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')`.

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.models import load_model
```

```

from PIL import Image

model = load_model(r'C:\Users\smerf\Downloads\Tugas6_B_11881\vgg-16.h5')
class_names = ['Merah', 'Thompson', 'Concord']

def classify_images(image_path, save_path='predicted_image.jpg'):
    try:
        input_image = tf.keras.utils.load_img(image_path,
        target_size=(180, 180))
        input_image_array = tf.keras.utils.img_to_array(input_image)
        input_image_exp_dim = tf.expand_dims(input_image_array, 0)

        predictions = model.predict(input_image_exp_dim)
        result = tf.nn.softmax(predictions[0])
        class_idx = np.argmax(result)
        confidence = np.max(result) * 100

        print(f"Prediksi: {class_names[class_idx]}")
        print(f"Confidence: {confidence:.2f}%")

        input_image = Image.open(image_path)
        input_image.save(save_path)

        return f"Prediksi: {class_names[class_idx]} dengan confidence
        {confidence:.2f}%. Gambar asli disimpan di {save_path}."
    except Exception as e:
        return f"Terjadi kesalahan: {e}"

result = classify_images(r'C:\Users\smerf\Downloads\Tugas6_B_11881\
train_data\Concord\Concord_Grape_Original_Data009.jpg',
save_path='concord9.jpg')
print(result)

```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

1/1 ————— 0s 406ms/step

Prediksi: Merah

Confidence: 33.49%

Prediksi: Merah dengan confidence 33.49%. Gambar asli disimpan di concord9.jpg.

```

import tensorflow as tf
from tensorflow.keras.models import load_model
import seaborn as sns
import matplotlib.pyplot as plt

```

```

test_data = tf.keras.preprocessing.image_dataset_from_directory(
    r'test_data',

```

```

        labels='inferred',
        label_mode='categorical',
        batch_size=32,
        image_size=(180, 180)
    )

y_pred = model.predict(test_data)
y_pred_class = tf.argmax(y_pred, axis=1)

true_labels = []
for _, labels in test_data:
    true_labels.extend(tf.argmax(labels, axis=1).numpy())
true_labels = tf.convert_to_tensor(true_labels)

conf_mat = tf.math.confusion_matrix(true_labels, y_pred_class)
accuracy = tf.reduce_sum(tf.linalg.diag_part(conf_mat)) /
tf.reduce_sum(conf_mat)

precision = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=0)
recall = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=1)

f1_score = 2 * (precision * recall) / (precision + recall)

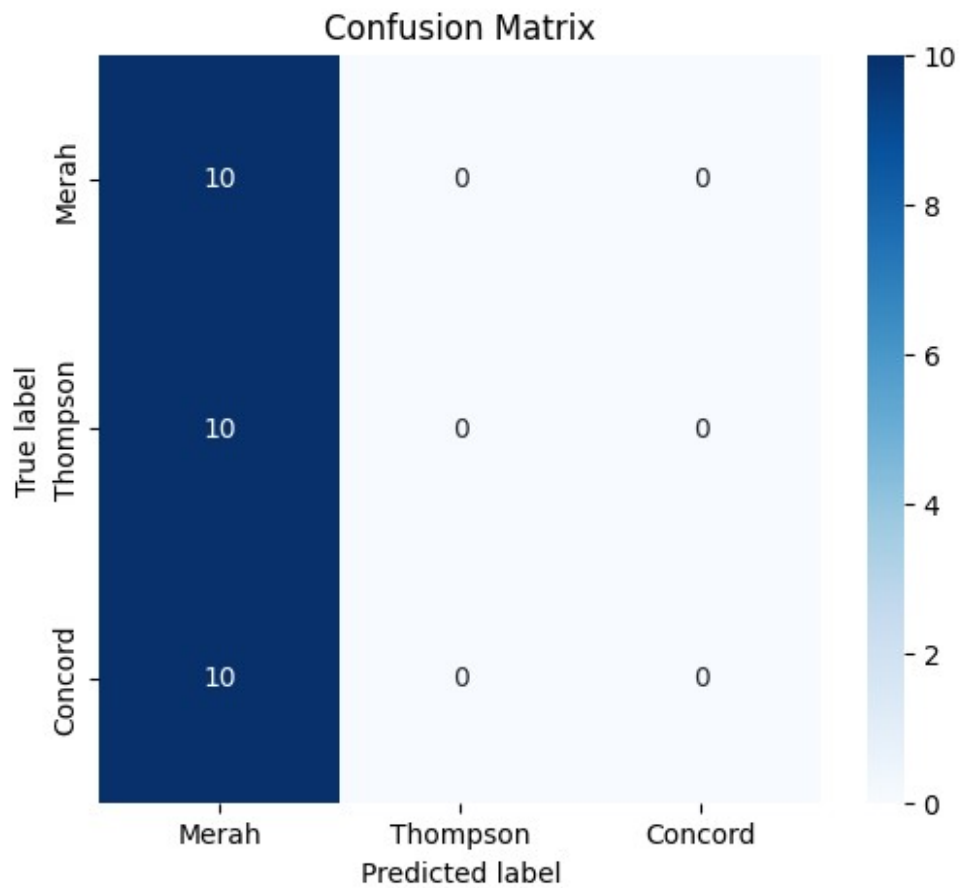
plt.figure(figsize=(6, 5))
sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
            xticklabels=["Merah", "Thompson", "Concord"],
            yticklabels=["Merah", "Thompson", "Concord"])
plt.title('Confusion Matrix')
plt.xlabel('Predicted label')
plt.ylabel('True label')
plt.show()

print("Confusion Matrix:\n", conf_mat.numpy())
print("Akurasi:", accuracy.numpy())
print("Presisi:", precision.numpy())
print("Recall:", recall.numpy())
print("F1 Score:", f1_score.numpy())

Found 30 files belonging to 3 classes.

1/1 ————— 4s 4s/step

```



Confusion Matrix:

```
[[10  0  0]
```

```
[10  0  0]
```

```
[10  0  0]]
```

Akurasi: 0.3333333333333333

Presisi: [0.33333333 nan nan]

Recall: [1. 0. 0.]

F1 Score: [0.5 nan nan]

Julius Aditya 220711892 Bokeh Mengklisifikasi Anggur Merah, Concord, dan Thompson

```
import tensorflow as tf
import cv2
import numpy as np
from matplotlib import pyplot as plt
data_dir = r"C:\Users\LENOVO\Videos\TUBESMLUAS\train_data"
data = tf.keras.utils.image_dataset_from_directory(data_dir, seed =
123, image_size=(180,180), batch_size=16)

print(data.class_names)

class_names = data.class_names

img_size = 180
batch = 32
validation_split = 0.1
dataset = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
)
total_count = len(dataset)
val_count = int(total_count * validation_split)
train_count = total_count - val_count

print("Total Images:", total_count)
print("Train Images:", train_count)
print("Validation Images:", val_count)

train_ds = dataset.take(train_count)
val_ds = dataset.skip(train_count)

Found 300 files belonging to 3 classes.
['Concord', 'Merah', 'Thompson']
Found 300 files belonging to 3 classes.
Total Images: 10
Train Images: 9
Validation Images: 1

import matplotlib.pyplot as plt

i = 0
plt.figure(figsize=(10,10))

for images, labels in train_ds.take(1):
    for i in range(9):
        plt.subplot(3,3, i+1)
        plt.imshow(images[i].numpy().astype('uint8'))
```

```
plt.title(class_names[labels[i]])
plt.axis('off')
```

Thompson



Concord



Thompson



Merah



Merah



Thompson



Thompson



Thompson



Merah



```
for images, labels in train_ds.take(1):
    images_array = np.array(images)
    print(images_array.shape)
```

```
(32, 180, 180, 3)
```

```
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential, load_model
```

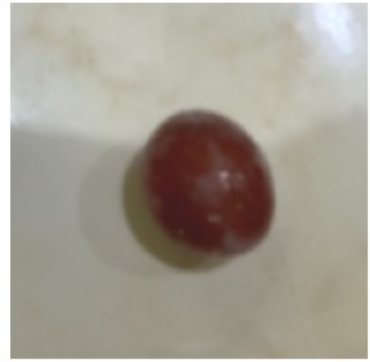
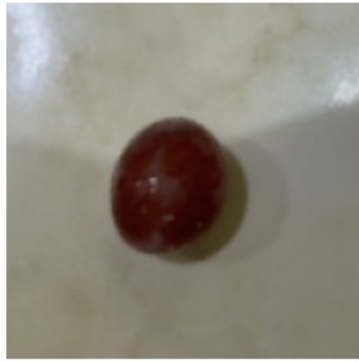
```

Tuner = tf.data.AUTOTUNE
train_ds = train_ds.cache().shuffle(1000).prefetch(buffer_size =
Tuner)
val_ds = val_ds.cache().shuffle(1000).prefetch(buffer_size = Tuner)

data_augmentation = Sequential([
    layers.RandomFlip("horizontal", input_shape=
(img_size,img_size,3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1),
    layers.RandomBrightness(0.1)
])

i = 0
plt.figure(figsize=(10,10))
for images, labels in train_ds.take(69):
    for i in range(9):
        images = data_augmentation(images)
        plt.subplot(3,3, i+1)
        plt.imshow(images[0].numpy().astype('uint8'))
        plt.axis('off')

```



```
import tensorflow as tf
from keras.models import Model
from keras.layers import Input, Conv2D, MaxPool2D, Flatten, Dense,
Dropout
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.preprocessing.image import ImageDataGenerator

def alexnet(input_shape, n_classes):
    input = Input(input_shape)

    x = Conv2D(96, (11, 11), strides=4, activation='relu',
```



```

padding='valid')(input)
    x = MaxPool2D((3, 3), strides=2)(x)

    x = Conv2D(256, (5, 5), activation='relu', padding='same')(x)
    x = MaxPool2D((3, 3), strides=2)(x)

    x = Conv2D(384, (3, 3), activation='relu', padding='same')(x)
    x = Conv2D(384, (3, 3), activation='relu', padding='same')(x)

    x = Conv2D(256, (3, 3), activation='relu', padding='same')(x)
    x = MaxPool2D((3, 3), strides=2)(x)

    x = Flatten()(x)

    x = Dense(4096, activation='relu')(x)
    x = Dropout(0.5)(x)
    x = Dense(4096, activation='relu')(x)
    x = Dropout(0.5)(x)
    x = Dense(1000, activation='relu')(x)

    output = Dense(n_classes, activation='softmax')(x)

    model = Model(input, output)
    return model

input_shape = (227, 227, 3)
n_classes = 3

model = alexnet(input_shape, n_classes)
model.summary()

model.compile(
    optimizer=Adam(),
    loss='sparse_categorical_crossentropy',
    metrics=['accuracy']
)

datagen = ImageDataGenerator(rescale=1./255)
val_datagen = ImageDataGenerator(rescale=1./255)

train_ds = datagen.flow_from_directory(
    r'C:\Users\LENOVO LEGION\Videos\TUBESMLUAS\train_data',
    target_size=(227, 227),
    batch_size=32,
    class_mode='sparse'
)

val_ds = val_datagen.flow_from_directory(
    r'C:\Users\LENOVO LEGION\Videos\TUBESMLUAS\train_data',

```

```

        target_size=(227, 227),
        batch_size=32,
        class_mode='sparse'
    )

    early_stopping = EarlyStopping(
        monitor='val_accuracy',
        patience=5,
        mode='max'
    )

    history = model.fit(
        train_ds,
        epochs=30,
        validation_data=val_ds,
        callbacks=[early_stopping],
        verbose=1
    )

```

Model: "functional_11"

Layer (type)	Output Shape
Param #	
input_layer_7 (InputLayer)	(None, 227, 227, 3)
0	
conv2d_15 (Conv2D)	(None, 55, 55, 96)
34,944	
max_pooling2d_9 (MaxPooling2D)	(None, 27, 27, 96)
0	
conv2d_16 (Conv2D)	(None, 27, 27, 256)
614,656	
max_pooling2d_10 (MaxPooling2D)	(None, 13, 13, 256)
0	
conv2d_17 (Conv2D)	(None, 13, 13, 384)
885,120	

conv2d_18 (Conv2D)	(None, 13, 13, 384)	
1,327,488		
conv2d_19 (Conv2D)	(None, 13, 13, 256)	
884,992		
max_pooling2d_11 (MaxPooling2D)	(None, 6, 6, 256)	
0		
flatten_3 (Flatten)	(None, 9216)	
0		
dense_12 (Dense)	(None, 4096)	
37,752,832		
dropout_6 (Dropout)	(None, 4096)	
0		
dense_13 (Dense)	(None, 4096)	
16,781,312		
dropout_7 (Dropout)	(None, 4096)	
0		
dense_14 (Dense)	(None, 1000)	
4,097,000		
dense_15 (Dense)	(None, 3)	
3,003		

Total params: 62,381,347 (237.97 MB)

Trainable params: 62,381,347 (237.97 MB)

Non-trainable params: 0 (0.00 B)

Found 300 images belonging to 3 classes.
Found 300 images belonging to 3 classes.

```
Epoch 1/30
10/10 _____ 82s 6s/step - accuracy: 0.2957 - loss:
1.2819 - val_accuracy: 0.3333 - val_loss: 1.0748
Epoch 2/30
10/10 _____ 74s 5s/step - accuracy: 0.4131 - loss:
0.9865 - val_accuracy: 0.6900 - val_loss: 0.6626
Epoch 3/30
10/10 _____ 74s 5s/step - accuracy: 0.6216 - loss:
0.6762 - val_accuracy: 0.7133 - val_loss: 0.5645
Epoch 4/30
10/10 _____ 76s 5s/step - accuracy: 0.6399 - loss:
0.7429 - val_accuracy: 0.4267 - val_loss: 1.1534
Epoch 5/30
10/10 _____ 72s 5s/step - accuracy: 0.3752 - loss:
1.2006 - val_accuracy: 0.3333 - val_loss: 1.1052
Epoch 6/30
10/10 _____ 77s 6s/step - accuracy: 0.3161 - loss:
1.1026 - val_accuracy: 0.5433 - val_loss: 0.9953
Epoch 7/30
10/10 _____ 79s 6s/step - accuracy: 0.3993 - loss:
1.1281 - val_accuracy: 0.3333 - val_loss: 1.0984
Epoch 8/30
10/10 _____ 73s 5s/step - accuracy: 0.2945 - loss:
1.1013 - val_accuracy: 0.3333 - val_loss: 1.0988
```

```
import matplotlib.pyplot as plt

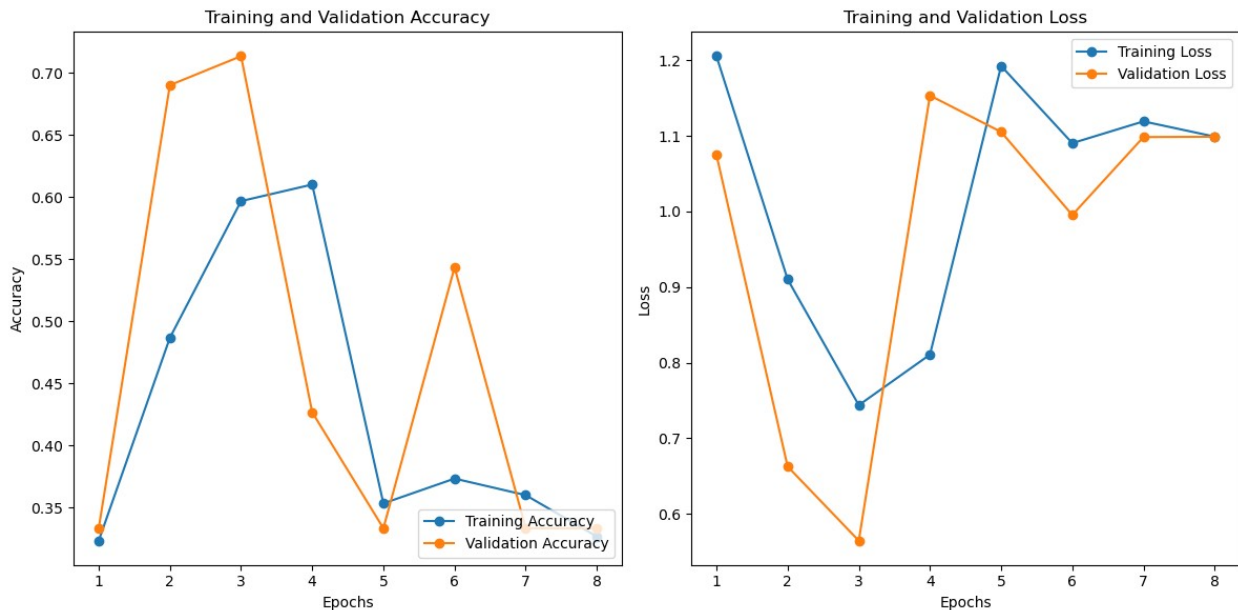
epochs_range = range(1, len(history.history['loss']) + 1)

plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 1)
plt.plot(epochs_range, history.history['accuracy'], label='Training
Accuracy', marker='o')
plt.plot(epochs_range, history.history['val_accuracy'],
label='Validation Accuracy', marker='o')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')

plt.subplot(1, 2, 2)
plt.plot(epochs_range, history.history['loss'], label='Training Loss',
marker='o')
plt.plot(epochs_range, history.history['val_loss'], label='Validation
Loss', marker='o')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
```

```
plt.tight_layout()
plt.show()
```



```
model.save('alexnet.h5')
```

c:\Users\LENOVO LEGION\anaconda3\Lib\site-packages\keras\src\models\model.py:342: UserWarning: You are saving your model as an HDF5 file via `model.save()`. This file format is considered legacy. We recommend using instead the native Keras format, e.g.

```
`model.save('my_model.keras')`.
```

```
warnings.warn()
```

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.models import load_model
from PIL import Image
```

```
model = load_model(r'C:\Users\LENOVO LEGION\Videos\TUBESMLUAS\alexnet.h5')
```

```
class_names = ['Concord', 'Merah', 'Thompson']
```

```
def classify_images(image_path, save_path='predicted_image.jpg'):
    try:
        input_image = tf.keras.utils.load_img(image_path,
        target_size=(180, 180))
        input_image_array = tf.keras.utils.img_to_array(input_image)
        input_image_exp_dim = tf.expand_dims(input_image_array, 0)

        predictions = model.predict(input_image_exp_dim)
```

```

        result = tf.nn.softmax(predictions[0])
        class_idx = np.argmax(result)
        confidence = np.max(result) * 100

        print(f"Prediksi: {class_names[class_idx]}")
        print(f"Confidence: {confidence:.2f}%")

        input_image = Image.open(image_path)
        input_image.save(save_path)

        return f"Prediksi: {class_names[class_idx]} dengan confidence
{confidence:.2f}%. Gambar asli disimpan di {save_path}."
    except Exception as e:
        return f"Terjadi kesalahan: {e}"

result = classify_images(r'test_data\Merah\
Merah_Grape_Original_Data03.JPG', save_path='merah2.jpg')
print(result)

```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

Terjadi kesalahan: Input 0 of layer "functional_11" is incompatible with the layer: expected shape=(None, 227, 227, 3), found shape=(1, 180, 180, 3)

```

import tensorflow as tf
from tensorflow.keras.models import load_model
import seaborn as sns
import matplotlib.pyplot as plt

test_data = tf.keras.preprocessing.image_dataset_from_directory(
    r'test_data',
    labels='inferred',
    label_mode='categorical',
    batch_size=32,
    image_size=(227, 227)
)

model = load_model(r'C:\Users\LENOVO LEGION\Videos\TUBESMLUAS\
alexnet.h5')

y_pred = model.predict(test_data)
y_pred_class = tf.argmax(y_pred, axis=1)

true_labels = []
for _, labels in test_data:
    true_labels.extend(tf.argmax(labels, axis=1).numpy())
true_labels = tf.convert_to_tensor(true_labels)

```

```

conf_mat = tf.math.confusion_matrix(true_labels, y_pred_class)

accuracy = tf.reduce_sum(tf.linalg.diag_part(conf_mat)) /
tf.reduce_sum(conf_mat)

precision = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=0)
recall = tf.linalg.diag_part(conf_mat) / tf.reduce_sum(conf_mat,
axis=1)

f1_score = 2 * (precision * recall) / (precision + recall)

plt.figure(figsize=(6, 5))
sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
            xticklabels=["Concord", "Merah", "Thompson"],
            yticklabels=["Concord", "Merah", "Thompson"])
plt.title('Confusion Matrix')
plt.xlabel('Predicted label')
plt.ylabel('True label')
plt.show()

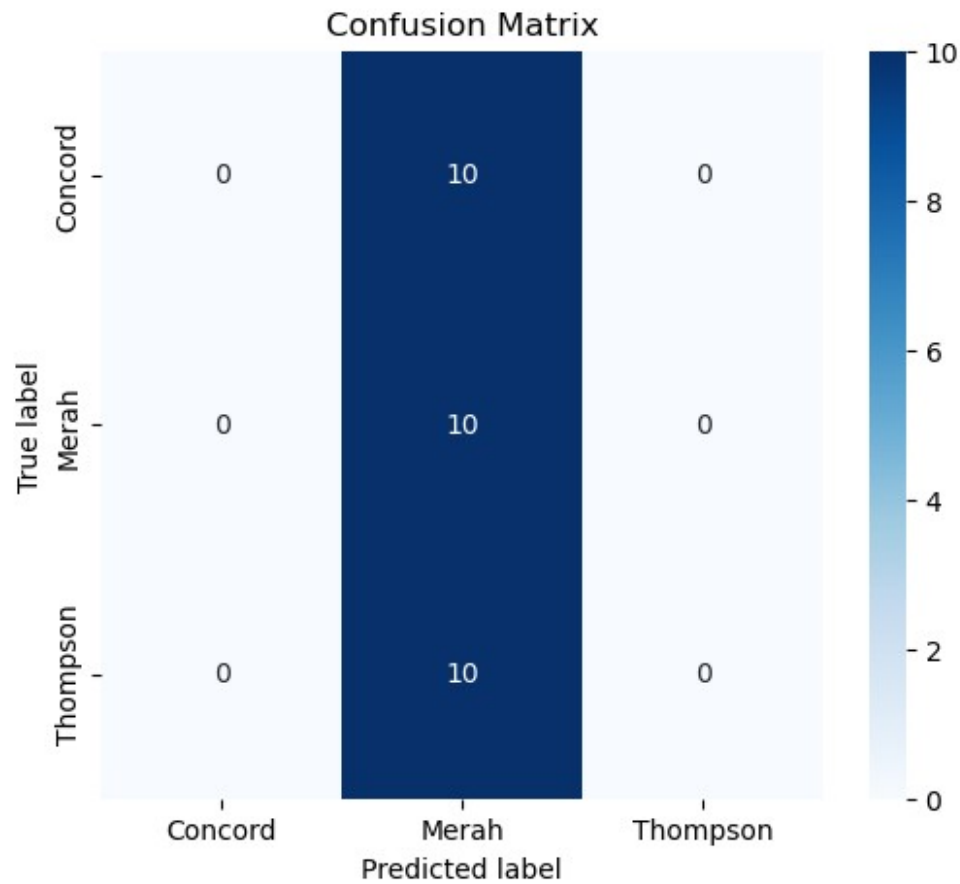
print("Confusion Matrix:\n", conf_mat.numpy())
print("Akurasi: ", accuracy.numpy())
print("Presisi: ", precision.numpy())
print("Recall:", recall.numpy())
print("F1 Score: ", f1_score.numpy())

```

Found 30 files belonging to 3 classes.

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or evaluate the model.

1/1 ————— 1s 827ms/step



```
Confusion Matrix:
[[ 0 10  0]
 [ 0 10  0]
 [ 0 10  0]]
Akurasi: 0.3333333333333333
Presisi: [ nan 0.33333333 nan]
Recall: [0. 1. 0.]
F1 Score: [nan 0.5 nan]
```