

# GOOGLE NET

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

# Load data
data_dir = r"D:\coolyeah\semester5\ml\tubes_uas\train_data\train_data"
img_size = 180
batch = 32

# Load dataset
dataset = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
)

# Total images
total_count = len(dataset) * batch # Hitung total gambar
print("Total Images: ", total_count)

# Calculate counts for train, validation, and test sets
train_count = int(total_count * 0.8)
val_count = int(total_count * 0.1)
test_count = total_count - train_count - val_count

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

# Split dataset into train, validation, and test sets
train_ds = dataset.take(train_count // batch)
val_ds = dataset.skip(train_count // batch).take(val_count // batch)
test_ds = dataset.skip(train_count // batch + val_count //
batch).take(test_count // batch)

# Check class names
class_names = dataset.class_names
print("Class Names: ", class_names)

Found 301 files belonging to 3 classes.
Total Images: 320
Train Images: 256
Validation Images: 32
Test Images: 32
Class Names: ['Busuk', 'Matang', 'Mentah']

import matplotlib.pyplot as plt
```

```

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

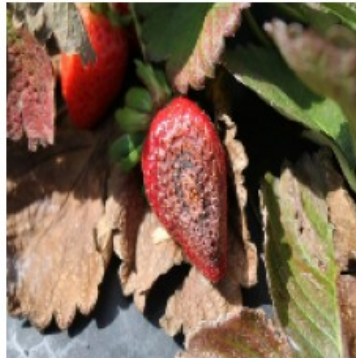
for images, labels in train_ds.take(1):
    for i in range(9):
        plt.subplot(3,3, i+1)
        plt.imshow(images[i].numpy().astype('uint8'))
        plt.title(class_names[labels[i]])
        plt.axis('off')

```

Mentah



Busuk



Matang



Mentah



Busuk



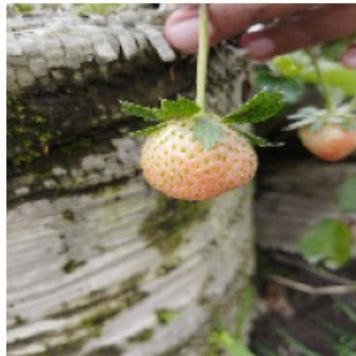
Mentah



Busuk



Mentah



Matang



```

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))

(32, 180, 180, 3)

import os
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers, Sequential
from PIL import Image
import matplotlib.pyplot as plt

# If using Jupyter Notebook, uncomment the following line
# %matplotlib inline

# Define image size and batch size
img_size = 180 # Example image size
batch_size = 32 # Example batch size

# Define the path to your dataset
data_dir = r"D:\coolyeah\semester5\ml\tubes_uas\train_data\train_data"
# Replace with your dataset path

# Function to load images and labels
def load_images_from_directory(directory):
    images = []
    labels = []
    class_names = os.listdir(directory) # Get class names from directory
    for label, class_name in enumerate(class_names):
        class_dir = os.path.join(directory, class_name)
        if os.path.isdir(class_dir):
            for file_name in os.listdir(class_dir):
                if file_name.lower().endswith(('.png', '.jpg', '.jpeg', '.gif', '.bmp')):
                    file_path = os.path.join(class_dir, file_name)
                    try:
                        img = Image.open(file_path).convert('RGB') # Open and convert to RGB
                        img = img.resize((img_size, img_size)) # Resize image
                        images.append(np.array(img)) # Convert to numpy array
                        labels.append(label) # Append label
                    except Exception as e:
                        print(f"Error loading image {file_path}: {e}")

```

```

    return np.array(images), np.array(labels)

# Load images and labels
images, labels = load_images_from_directory(data_dir)

# Check the shape of the loaded images
print(f"Loaded {len(images)} images with shape: {images.shape}")

# Create a TensorFlow dataset
train_ds = tf.data.Dataset.from_tensor_slices((images, labels))

# Shuffle and batch the dataset
Tuner = tf.data.AUTOTUNE
train_ds =
train_ds.shuffle(buffer_size=1000).batch(batch_size).prefetch(buffer_s
ize=Tuner)

# Data augmentation using Sequential
data_augmentation = Sequential([
    layers.RandomFlip("horizontal", input_shape=(img_size, img_size,
3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1)
])

# Visualize augmented images
plt.figure(figsize=(10, 10))
for images_batch, labels_batch in train_ds.take(1):
    augmented_images = data_augmentation(images_batch) # Apply data
augmentation
    for i in range(min(9, augmented_images.shape[0])): # Ensure we
don't exceed the number of images
        plt.subplot(3, 3, i + 1)
        plt.imshow(augmented_images[i].numpy().astype('uint8'))
        plt.axis('off')
plt.show()

```

Loaded 301 images with shape: (301, 180, 180, 3)

```

c:\Users\capsl\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.

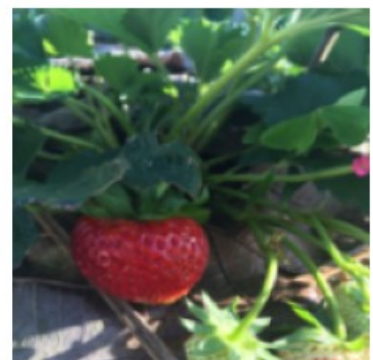
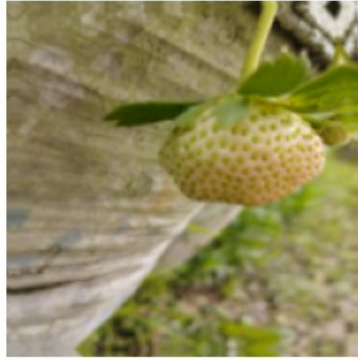
```

```

    super().__init__(**kwargs)

```





```
import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.layers import Input, Dense, Conv2D, Flatten,
MaxPooling2D, AveragePooling2D, Dropout, BatchNormalization
from tensorflow.keras import regularizers
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.callbacks import EarlyStopping,
ReduceLR0nPlateau

def googlenet(input_shape, n_classes):
    def inception_block(x, filters):
```

```

    # 1x1 Convolution
    branch1x1 = Conv2D(filters[0], (1, 1), padding='same',
activation='relu')(x)
    branch1x1 = BatchNormalization()(branch1x1)

    # 1x1 Convolution followed by 3x3 Convolution
    branch3x3 = Conv2D(filters[1], (1, 1), padding='same',
activation='relu')(x)
    branch3x3 = BatchNormalization()(branch3x3)
    branch3x3 = Conv2D(filters[2], (3, 3), padding='same',
activation='relu')(branch3x3)
    branch3x3 = BatchNormalization()(branch3x3)

    # 1x1 Convolution followed by 5x5 Convolution
    branch5x5 = Conv2D(filters[3], (1, 1), padding='same',
activation='relu')(x)
    branch5x5 = BatchNormalization()(branch5x5)
    branch5x5 = Conv2D(filters[4], (5, 5), padding='same',
activation='relu')(branch5x5)
    branch5x5 = BatchNormalization()(branch5x5)

    # 3x3 MaxPooling followed by 1x1 Convolution
    branch_pool = MaxPooling2D((3, 3), strides=(1, 1),
padding='same')(x)
    branch_pool = Conv2D(filters[5], (1, 1), padding='same',
activation='relu')(branch_pool)
    branch_pool = BatchNormalization()(branch_pool)

    # Concatenate all branches
    outputs = layers.concatenate([branch1x1, branch3x3, branch5x5,
branch_pool], axis=-1)
    return outputs

input = Input(shape=input_shape)

# Initial Convolution Layer
x = Conv2D(64, (7, 7), strides=(2, 2), padding='same',
activation='relu')(input)
x = BatchNormalization()(x)
x = MaxPooling2D((3, 3), strides=(2, 2))(x)

# Inception Blocks
x = inception_block(x, [64, 128, 128, 32, 32, 32])
x = inception_block(x, [128, 128, 192, 96, 96, 64])
x = MaxPooling2D((3, 3), strides=(2, 2))(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])

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x = inception_block(x, [256, 160, 320, 32, 128, 128])
x = MaxPooling2D((3, 3), strides=(2, 2))(x)

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

# Average Pooling and Fully Connected Layer
x = AveragePooling2D((7, 7))(x)
x = Flatten()(x)
x = Dense(256, activation='relu',
kernel_regularizer=regularizers.l2(0.01))(x)
x = Dropout(0.5)(x)
output = Dense(n_classes, activation='softmax')(x)

model = models.Model(inputs=input, outputs=output)
return model

# Data Augmentation
datagen = ImageDataGenerator(
    rotation_range=40,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest'
)

# Input shape and class names
input_shape = (180, 180, 3)
class_names = ['Busuk', 'Matang', 'Mentah']
n_classes = len(class_names)

# Clear Keras session
tf.keras.backend.clear_session()

# Create model
model = googlenet(input_shape, n_classes)
model.summary()

Model: "functional"

```

Layer (type)	Output Shape	Param #	Connected to
input_layer	(None, 180, 180,	0	-

(InputLayer)	3)		
conv2d (Conv2D) input_layer[0][0]	(None, 90, 90, 64)	9,472	
batch_normalization (BatchNormalizatio...	(None, 90, 90, 64)	256	conv2d[0][0]
max_pooling2d batch_normalizat...	(None, 44, 44, 64)	0	
conv2d_2 (Conv2D) max_pooling2d[0]...	(None, 44, 44, 128)	8,320	
conv2d_4 (Conv2D) max_pooling2d[0]...	(None, 44, 44, 32)	2,080	
batch_normalization... [0]	(None, 44, 44, 128)	512	conv2d_2[0]
batch_normalization... [0]	(None, 44, 44, 32)	128	conv2d_4[0]
max_pooling2d_1 max_pooling2d[0]...	(None, 44, 44, 64)	0	



conv2d_1 (Conv2D)	(None, 44, 44,	4,160		
max_pooling2d[0]...	64)			
conv2d_3 (Conv2D)	(None, 44, 44,	147,584		
batch_normalizat...	128)			
conv2d_5 (Conv2D)	(None, 44, 44,	25,632		
batch_normalizat...	32)			
conv2d_6 (Conv2D)	(None, 44, 44,	2,080		
max_pooling2d_1[...	32)			
batch_normalizatio...	(None, 44, 44,	256	conv2d_1[0]	
[0]	64)			
(BatchNormalizatio...				
batch_normalizatio...	(None, 44, 44,	512	conv2d_3[0]	
[0]	128)			
(BatchNormalizatio...				
batch_normalizatio...	(None, 44, 44,	128	conv2d_5[0]	
[0]	32)			
(BatchNormalizatio...				
batch_normalizatio...	(None, 44, 44,	128	conv2d_6[0]	
[0]	32)			
(BatchNormalizatio...				

concatenate	(None, 44, 44,	0	
batch_normalizat...	(Concatenate)	256)	
batch_normalizat...			
batch_normalizat...			
batch_normalizat...			
conv2d_8 (Conv2D)	(None, 44, 44,	32,896	
concatenate[0][0]	128)		
conv2d_10 (Conv2D)	(None, 44, 44,	24,672	
concatenate[0][0]	96)		
batch_normalizatio...	(None, 44, 44,	512	conv2d_8[0]
[0]	(BatchNormalizatio...	128)	
batch_normalizatio...	(None, 44, 44,	384	conv2d_10[0]
[0]	(BatchNormalizatio...	96)	
max_pooling2d_2	(None, 44, 44,	0	
concatenate[0][0]	256)		
(MaxPooling2D)			
conv2d_7 (Conv2D)	(None, 44, 44,	32,896	
concatenate[0][0]	128)		
conv2d_9 (Conv2D)	(None, 44, 44,	221,376	

batch_normalizat...		192)		
conv2d_11 (Conv2D)	(None, 44, 44,	230,496		
batch_normalizat...		96)		
conv2d_12 (Conv2D)	(None, 44, 44,	16,448		
max_pooling2d_2[...		64)		
batch_normalizatio...	(None, 44, 44,	512	conv2d_7[0]	
[0]	(BatchNormalizatio...	128)		
batch_normalizatio...	(None, 44, 44,	768	conv2d_9[0]	
[0]	(BatchNormalizatio...	192)		
batch_normalizatio...	(None, 44, 44,	384	conv2d_11[0]	
[0]	(BatchNormalizatio...	96)		
batch_normalizatio...	(None, 44, 44,	256	conv2d_12[0]	
[0]	(BatchNormalizatio...	64)		
concatenate_1	(None, 44, 44,	0		
batch_normalizat...				
(Concatenate)	480)			
batch_normalizat...				
batch_normalizat...				
batch_normalizat...				

max_pooling2d_3	(None, 21, 21,	0		
concatenate_1[0]...	(MaxPooling2D)	480)		
conv2d_14 (Conv2D)	(None, 21, 21,	46,176		
max_pooling2d_3[...	96)			
conv2d_16 (Conv2D)	(None, 21, 21,	7,696		
max_pooling2d_3[...	16)			
batch_normalizatio...	(None, 21, 21,	384	conv2d_14[0]	
[0]	(BatchNormalizatio...	96)		
batch_normalizatio...	(None, 21, 21,	64	conv2d_16[0]	
[0]	(BatchNormalizatio...	16)		
max_pooling2d_4	(None, 21, 21,	0		
max_pooling2d_3[...	(MaxPooling2D)	480)		
conv2d_13 (Conv2D)	(None, 21, 21,	92,352		
max_pooling2d_3[...	192)			
conv2d_15 (Conv2D)	(None, 21, 21,	179,920		
batch_normalizat...	208)			

conv2d_17 (Conv2D)	(None, 21, 21,	19,248	
batch_normalizat...	48)		
conv2d_18 (Conv2D)	(None, 21, 21,	30,784	
max_pooling2d_4[...	64)		
batch_normalizatio...	(None, 21, 21,	768	conv2d_13[0]
[0]	(BatchNormalizatio...	192)	
batch_normalizatio...	(None, 21, 21,	832	conv2d_15[0]
[0]	(BatchNormalizatio...	208)	
batch_normalizatio...	(None, 21, 21,	192	conv2d_17[0]
[0]	(BatchNormalizatio...	48)	
batch_normalizatio...	(None, 21, 21,	256	conv2d_18[0]
[0]	(BatchNormalizatio...	64)	
concatenate_2	(None, 21, 21,	0	
batch_normalizat...			
(Concatenate)	512)		
batch_normalizat...			
batch_normalizat...			
batch_normalizat...			
conv2d_20 (Conv2D)	(None, 21, 21,	57,456	
concatenate_2[0]...			

	112)		
conv2d_22 (Conv2D)	(None, 21, 21,	12,312	
concatenate_2[0]...	24)		
batch_normalizatio...	(None, 21, 21,	448	conv2d_20[0]
[0]	(BatchNormalizatio...	112)	
batch_normalizatio...	(None, 21, 21,	96	conv2d_22[0]
[0]	(BatchNormalizatio...	24)	
max_pooling2d_5	(None, 21, 21,	0	
concatenate_2[0]...	(MaxPooling2D)	512)	
conv2d_19 (Conv2D)	(None, 21, 21,	82,080	
concatenate_2[0]...	160)		
conv2d_21 (Conv2D)	(None, 21, 21,	226,016	
batch_normalizat...	224)		
conv2d_23 (Conv2D)	(None, 21, 21,	38,464	
batch_normalizat...	64)		
conv2d_24 (Conv2D)	(None, 21, 21,	32,832	
max_pooling2d_5[...	64)		



batch_normalization_19[0]	(None, 21, 21, 160)	640	conv2d_19[0]
batch_normalization_21[0]	(None, 21, 21, 224)	896	conv2d_21[0]
batch_normalization_23[0]	(None, 21, 21, 64)	256	conv2d_23[0]
batch_normalization_24[0]	(None, 21, 21, 64)	256	conv2d_24[0]
concatenate_3 batch_normalization_25 (Concatenate) batch_normalization_26 batch_normalization_27 batch_normalization_28	(None, 21, 21, 512)	0	
conv2d_26 (Conv2D) concatenate_3[0]...	(None, 21, 21, 128)	65,664	
conv2d_28 (Conv2D) concatenate_3[0]...	(None, 21, 21, 24)	12,312	

batch_normalization_26[0]	(None, 21, 21, 128)	512	conv2d_26[0]
batch_normalization_28[0]	(None, 21, 21, 24)	96	conv2d_28[0]
max_pooling2d_6 concatenate_3[0]	(None, 21, 21, 512)	0	
conv2d_25 (Conv2D) concatenate_3[0]	(None, 21, 21, 128)	65,664	
conv2d_27 (Conv2D) batch_normalization_27	(None, 21, 21, 256)	295,168	
conv2d_29 (Conv2D) batch_normalization_29	(None, 21, 21, 64)	38,464	
conv2d_30 (Conv2D) max_pooling2d_6	(None, 21, 21, 64)	32,832	
batch_normalization_25[0]	(None, 21, 21, 128)	512	conv2d_25[0]
batch_normalization_27	(None, 21, 21, 1,024)	1,024	conv2d_27[0]

[0]	(BatchNormalization...	256)		
	batch_normalization...	(None, 21, 21,	256	conv2d_29[0]
[0]	(BatchNormalization...	64)		
	batch_normalization...	(None, 21, 21,	256	conv2d_30[0]
[0]	(BatchNormalization...	64)		
	concatenate_4	(None, 21, 21,	0	
batch_normalizat...	(Concatenate)	512)		
batch_normalizat...				
batch_normalizat...				
batch_normalizat...				
	conv2d_32 (Conv2D)	(None, 21, 21,	73,872	
concatenate_4[0]...		144)		
	conv2d_34 (Conv2D)	(None, 21, 21,	16,416	
concatenate_4[0]...		32)		
	batch_normalization...	(None, 21, 21,	576	conv2d_32[0]
[0]	(BatchNormalization...	144)		
	batch_normalization...	(None, 21, 21,	128	conv2d_34[0]
[0]	(BatchNormalization...	32)		

max_pooling2d_7 concatenate_4[0]...	(None, 21, 21,	0	
(MaxPooling2D)	512)		
conv2d_31 (Conv2D) concatenate_4[0]...	(None, 21, 21,	57,456	
	112)		
conv2d_33 (Conv2D) batch_normalizat...	(None, 21, 21,	373,536	
	288)		
conv2d_35 (Conv2D) batch_normalizat...	(None, 21, 21,	51,264	
	64)		
conv2d_36 (Conv2D) max_pooling2d_7[...	(None, 21, 21,	32,832	
	64)		
batch_normalizatio... [0]	(None, 21, 21,	448	conv2d_31[0]
(BatchNormalizatio...	112)		
batch_normalizatio... [0]	(None, 21, 21,	1,152	conv2d_33[0]
(BatchNormalizatio...	288)		
batch_normalizatio... [0]	(None, 21, 21,	256	conv2d_35[0]
(BatchNormalizatio...	64)		

batch_normalization_36[0]	(None, 21, 21, 64)	256	conv2d_36[0]
concatenate_5	(None, 21, 21, 528)	0	
batch_normalization_37			
batch_normalization_38			
batch_normalization_39			
conv2d_38 (Conv2D)	(None, 21, 21, 160)	84,640	
concatenate_5[0]			
conv2d_40 (Conv2D)	(None, 21, 21, 32)	16,928	
concatenate_5[0]			
batch_normalization_40[0]	(None, 21, 21, 160)	640	conv2d_38[0]
batch_normalization_41			
batch_normalization_42[0]	(None, 21, 21, 32)	128	conv2d_40[0]
max_pooling2d_8	(None, 21, 21, 528)	0	
concatenate_5[0]			
conv2d_37 (Conv2D)	(None, 21, 21, 256)	135,424	

conv2d_39 (Conv2D)	(None, 21, 21,	461,120		
batch_normalizat...	320)			
conv2d_41 (Conv2D)	(None, 21, 21,	102,528		
batch_normalizat...	128)			
conv2d_42 (Conv2D)	(None, 21, 21,	67,712		
max_pooling2d_8[...	128)			
batch_normalizatio...	(None, 21, 21,	1,024	conv2d_37[0]	
[0]	(BatchNormalizatio...	256)		
batch_normalizatio...	(None, 21, 21,	1,280	conv2d_39[0]	
[0]	(BatchNormalizatio...	320)		
batch_normalizatio...	(None, 21, 21,	512	conv2d_41[0]	
[0]	(BatchNormalizatio...	128)		
batch_normalizatio...	(None, 21, 21,	512	conv2d_42[0]	
[0]	(BatchNormalizatio...	128)		
concatenate_6	(None, 21, 21,	0		
batch_normalizat...	832)			
(Concatenate)				
batch_normalizat...				



batch_normalizat...			
batch_normalizat...			
max_pooling2d_9 concatenate_6[0]...	(None, 10, 10,	0	
(MaxPooling2D)	832)		
conv2d_44 (Conv2D) max_pooling2d_9[...	(None, 10, 10,	133,280	
	160)		
conv2d_46 (Conv2D) max_pooling2d_9[...	(None, 10, 10,	26,656	
	32)		
batch_normalizatio... [0]	(None, 10, 10,	640	conv2d_44[0]
(BatchNormalizatio...	160)		
batch_normalizatio... [0]	(None, 10, 10,	128	conv2d_46[0]
(BatchNormalizatio...	32)		
max_pooling2d_10 max_pooling2d_9[...	(None, 10, 10,	0	
(MaxPooling2D)	832)		
conv2d_43 (Conv2D) max_pooling2d_9[...	(None, 10, 10,	213,248	
	256)		
conv2d_45 (Conv2D)	(None, 10, 10,	461,120	

batch_normalizat...		320)		
conv2d_47 (Conv2D)	(None, 10, 10,	102,528		
batch_normalizat...		128)		
conv2d_48 (Conv2D)	(None, 10, 10,	106,624		
max_pooling2d_10...		128)		
batch_normalizatio...	(None, 10, 10,	1,024	conv2d_43[0]	
[0]	(BatchNormalizatio...	256)		
batch_normalizatio...	(None, 10, 10,	1,280	conv2d_45[0]	
[0]	(BatchNormalizatio...	320)		
batch_normalizatio...	(None, 10, 10,	512	conv2d_47[0]	
[0]	(BatchNormalizatio...	128)		
batch_normalizatio...	(None, 10, 10,	512	conv2d_48[0]	
[0]	(BatchNormalizatio...	128)		
concatenate_7	(None, 10, 10,	0		
batch_normalizat...				
(Concatenate)	832)			
batch_normalizat...				
batch_normalizat...				
batch_normalizat...				

conv2d_50 (Conv2D)	(None, 10, 10,	159,936	
concatenate_7[0]...	192)		
conv2d_52 (Conv2D)	(None, 10, 10,	39,984	
concatenate_7[0]...	48)		
batch_normalizatio...	(None, 10, 10,	768	conv2d_50[0]
[0]			
(BatchNormalizatio...	192)		
batch_normalizatio...	(None, 10, 10,	192	conv2d_52[0]
[0]			
(BatchNormalizatio...	48)		
max_pooling2d_11	(None, 10, 10,	0	
concatenate_7[0]...			
(MaxPooling2D)	832)		
conv2d_49 (Conv2D)	(None, 10, 10,	319,872	
concatenate_7[0]...			
	384)		
conv2d_51 (Conv2D)	(None, 10, 10,	663,936	
batch_normalizat...			
	384)		
conv2d_53 (Conv2D)	(None, 10, 10,	153,728	
batch_normalizat...			
	128)		

conv2d_54 (Conv2D)	(None, 10, 10,	106,624	
max_pooling2d_11...	128)		
batch_normalizatio...	(None, 10, 10,	1,536	conv2d_49[0]
[0]	(BatchNormalizatio...	384)	
batch_normalizatio...	(None, 10, 10,	1,536	conv2d_51[0]
[0]	(BatchNormalizatio...	384)	
batch_normalizatio...	(None, 10, 10,	512	conv2d_53[0]
[0]	(BatchNormalizatio...	128)	
batch_normalizatio...	(None, 10, 10,	512	conv2d_54[0]
[0]	(BatchNormalizatio...	128)	
concatenate_8	(None, 10, 10,	0	
batch_normalizat...			
(Concatenate)	1024)		
batch_normalizat...			
batch_normalizat...			
batch_normalizat...			
average_pooling2d	(None, 1, 1,	0	
concatenate_8[0]...			
(AveragePooling2D)	1024)		
flatten (Flatten)	(None, 1024)	0	
average_pooling2...			

dense (Dense)	(None, 256)	262,400	flatten[0][0]
dropout (Dropout)	(None, 256)	0	dense[0][0]
dense_1 (Dense)	(None, 3)	771	dropout[0][0]

Total params: 6,346,531 (24.21 MB)

Trainable params: 6,332,259 (24.16 MB)

Non-trainable params: 14,272 (55.75 KB)

```
import os
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.optimizers import Adam
from PIL import Image
import matplotlib.pyplot as plt

# Define image size and batch size
img_size = 180 # Ukuran gambar
batch_size = 32 # Ukuran batch
num_classes = 10 # Ganti dengan jumlah kelas yang sesuai

# Define the path to your dataset
data_dir = r"D:\coolyeah\semester5\ml\tubes_uas\train_data\train_data"
# Ganti dengan path dataset Anda

# Function to load images and labels
def load_images_from_directory(directory):
    images = []
    labels = []
    class_names = os.listdir(directory) # Mendapatkan nama kelas dari
    direktori
    for label, class_name in enumerate(class_names):
        class_dir = os.path.join(directory, class_name)
        if os.path.isdir(class_dir):
            for file_name in os.listdir(class_dir):
                if file_name.lower().endswith(('.png', '.jpg',
```

```

'.jpeg', '.gif', '.bmp')):
    file_path = os.path.join(class_dir, file_name)
    try:
        img = Image.open(file_path).convert('RGB') #
        Buka dan konversi ke RGB
        img = img.resize((img_size, img_size)) # Ubah
        ukuran gambar ke (180, 180)
        images.append(np.array(img)) # Konversi ke
        numpy array
        labels.append(label) # Tambahkan label
    except Exception as e:
        print(f"Error loading image {file_path}: {e}")
    return np.array(images), np.array(labels)

# Load training images and labels
images, labels = load_images_from_directory(data_dir)

# Create a TensorFlow dataset for training
train_ds = tf.data.Dataset.from_tensor_slices((images, labels)) #
Gunakan labels asli
train_ds =
train_ds.shuffle(buffer_size=1000).batch(batch_size).prefetch(tf.data.
AUTOTUNE)

# Define a simple CNN model
model = tf.keras.Sequential([
    layers.Input(shape=(img_size, img_size, 3)), # Input layer untuk
    gambar
    layers.Conv2D(32, (3, 3), activation='relu'),
    layers.MaxPooling2D(pool_size=(2, 2)),
    layers.Conv2D(64, (3, 3), activation='relu'),
    layers.MaxPooling2D(pool_size=(2, 2)),
    layers.Flatten(), # Flatten the output
    layers.Dense(512, activation='relu'), # Dense layer
    layers.Dense(num_classes, activation='softmax') # Output layer
    untuk num_classes
])

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

# Buat early stopping
early_stopping = EarlyStopping(monitor='accuracy', patience=5,
mode='max')

```



```

# Fit model tanpa validation_data
history = model.fit(train_ds,
                    epochs=30,
                    callbacks=[early_stopping])

# Buat plot dengan menggunakan history supaya jumlahnya sesuai epoch
yang dilakukan
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.legend(loc='lower right')
plt.title('Training Accuracy')

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

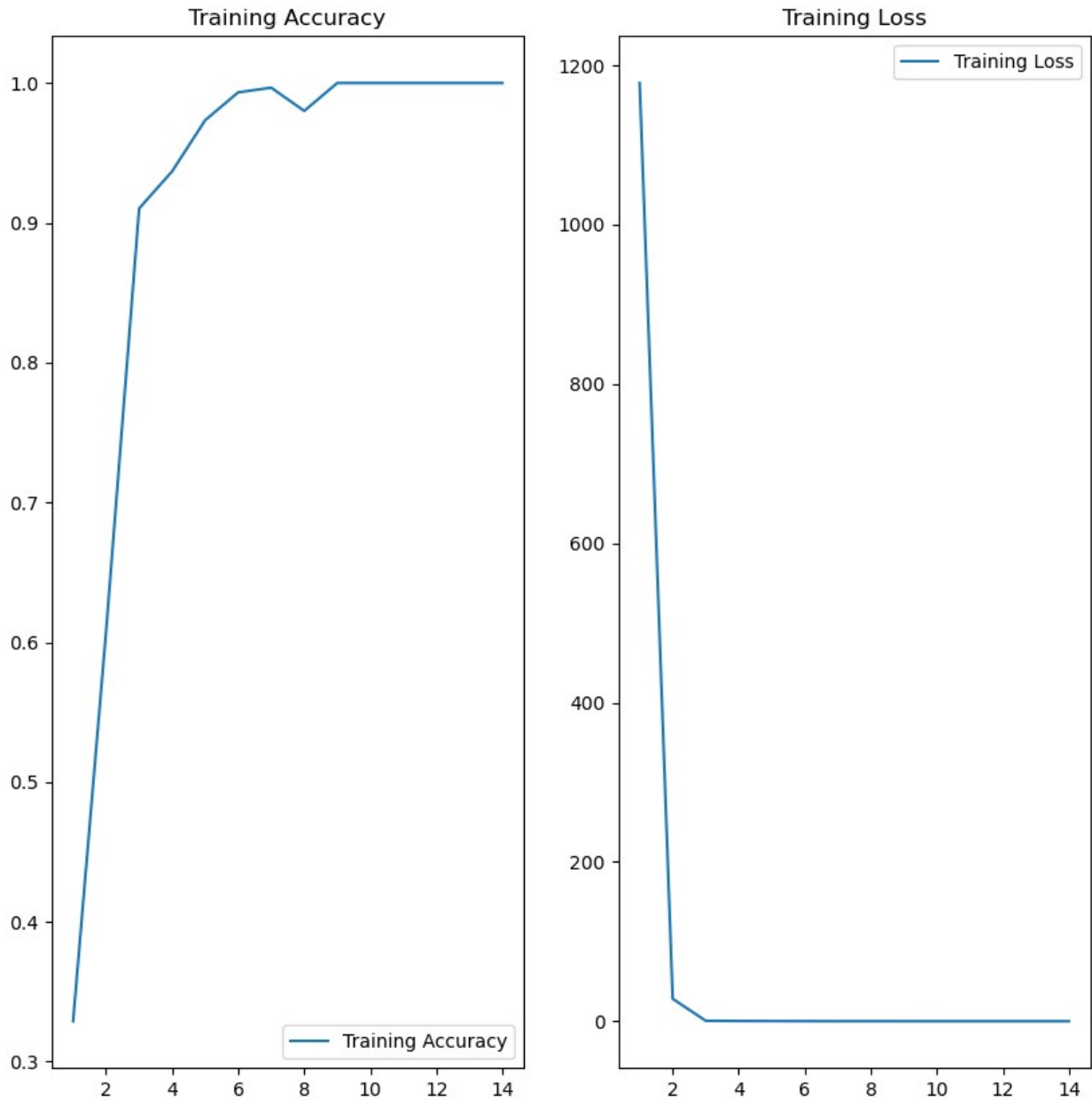
```

```

Epoch 1/30
10/10 _____ 7s 506ms/step - accuracy: 0.2643 - loss:
1206.0455
Epoch 2/30
10/10 _____ 5s 497ms/step - accuracy: 0.5002 - loss:
60.9073
Epoch 3/30
10/10 _____ 5s 503ms/step - accuracy: 0.8891 - loss:
0.8052
Epoch 4/30
10/10 _____ 5s 495ms/step - accuracy: 0.9375 - loss:
0.2679
Epoch 5/30
10/10 _____ 5s 505ms/step - accuracy: 0.9636 - loss:
0.2360
Epoch 6/30
10/10 _____ 5s 490ms/step - accuracy: 0.9962 - loss:
0.0685
Epoch 7/30
10/10 _____ 5s 499ms/step - accuracy: 0.9991 - loss:
0.0143
Epoch 8/30
10/10 _____ 5s 502ms/step - accuracy: 0.9794 - loss:
0.0800
Epoch 9/30
10/10 _____ 5s 498ms/step - accuracy: 1.0000 - loss:
0.0292
Epoch 10/30
10/10 _____ 5s 500ms/step - accuracy: 1.0000 - loss:

```

```
0.0104
Epoch 11/30
10/10 _____ 5s 499ms/step - accuracy: 1.0000 - loss:
0.0121
Epoch 12/30
10/10 _____ 5s 499ms/step - accuracy: 1.0000 - loss:
0.0019
Epoch 13/30
10/10 _____ 5s 499ms/step - accuracy: 1.0000 - loss:
0.0019
Epoch 14/30
10/10 _____ 5s 495ms/step - accuracy: 1.0000 - loss:
0.0026
```



```
model.save('gNet5.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'D:\coolyeah\semester5\ml\tubes_uas\
BestModel_GoogleNet_Matplotlib.h5') # Ganti dengan path model Anda
class_names = ['Busuk', 'Matang', 'Mentah']

# 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'D:\coolyeah\semester5\ml\tubes_uas\
test_data\test_data\Busuk\busuk_02.jpg', save_path='busuk_01.jpg')
result = classify_images(r'D:\coolyeah\semester5\ml\tubes_uas\
test_data\test_data\Matang\matang_05.jpg', save_path='matang_01.jpg')
result = classify_images(r'D:\coolyeah\semester5\ml\tubes_uas\
test_data\test_data\Mentah\mentah_04.jpg', save_path='mentah_01.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:5 out of the last 11 calls to <function TensorFlowTrainer.make\_predict\_function.<locals>.one\_step\_on\_data\_distributed at 0x00000211B144D580> 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](https://www.tensorflow.org/guide/function#controlling_retracing) and [https://www.tensorflow.org/api\\_docs/python/tf/function](https://www.tensorflow.org/api_docs/python/tf/function) for more details.

WARNING:tensorflow:5 out of the last 11 calls to <function TensorFlowTrainer.make\_predict\_function.<locals>.one\_step\_on\_data\_distributed at 0x00000211B144D580> 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](https://www.tensorflow.org/guide/function#controlling_retracing) and [https://www.tensorflow.org/api\\_docs/python/tf/function](https://www.tensorflow.org/api_docs/python/tf/function) for more details.

1/1 ————— 0s 94ms/step

Prediksi: Busuk

Confidence: 23.20%

1/1 ————— 0s 46ms/step

Prediksi: Matang

Confidence: 23.20%

1/1 ————— 0s 26ms/step

Prediksi: Mentah

Confidence: 23.20%

Prediksi: Mentah dengan confidence 23.20%. Gambar asli disimpan di mentah\_01.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
    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)

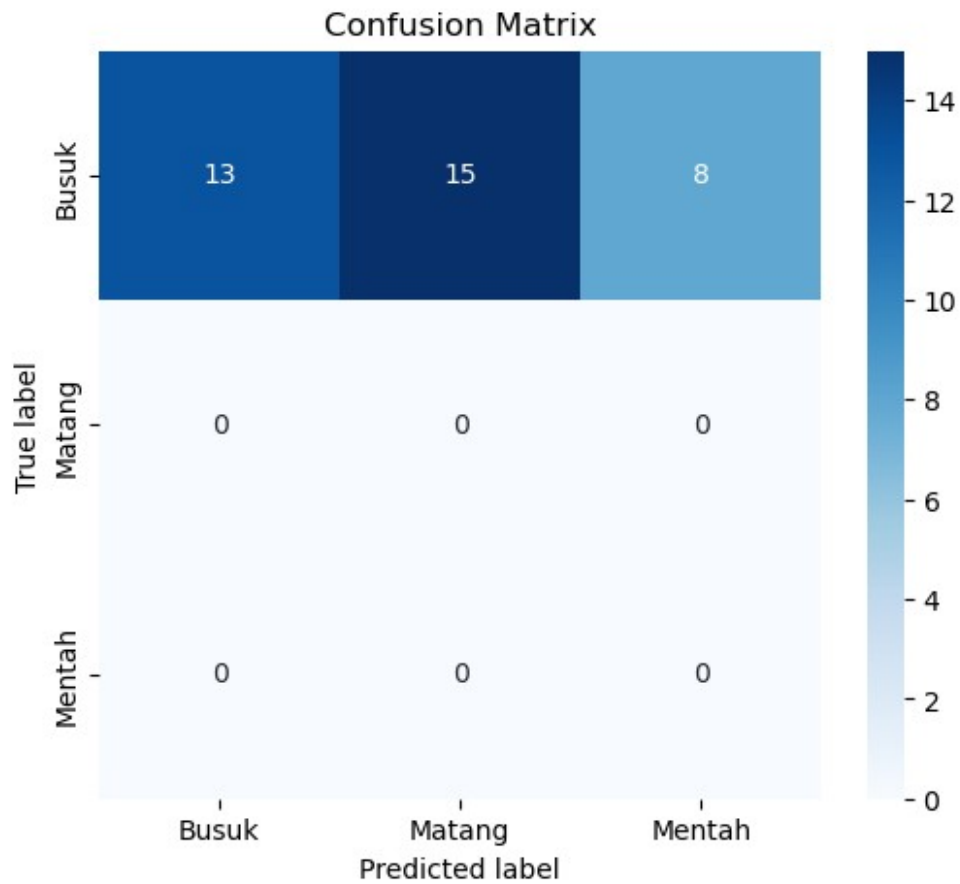
    # Visualisasi Confusion Matrix
    plt.figure(figsize=(6, 5))
    sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
                xticklabels=["Busuk", "Matang", "Mentah"],
                yticklabels=["Busuk", "Matang", "Mentah"])
    plt.title('Confusion Matrix')
    plt.xlabel('Predicted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

    # Menampilkan hasil
    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 36 files belonging to 1 classes.
    2/2                      0s 93ms/step

```





```

Confusion Matrix:
[[13 15  8]
 [ 0  0  0]
 [ 0  0  0]]
Akurasi:  0.3611111111111111
Presisi:  [1.  0.  0.]
Recall:   [0.36111111 nan nan]
F1 Score: [0.53061224 nan nan]

```

# MOBILE NET

```
import tensorflow as tf
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion_matrix, classification_report
import os
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dense,
Flatten, Dropout
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
from tensorflow.keras import models, layers
```

```
data_dir = r'/Users/saktiyoga/Development/PMDPM/train_data'
```

```
def load_data(data_dir, img_size=(224, 224), batch_size=32,
augment=False):
```

```
    if augment:
```

```
        data_gen = ImageDataGenerator(
            rescale=1./255,
            rotation_range=30,
            width_shift_range=0.2,
            height_shift_range=0.2,
            shear_range=0.2,
            zoom_range=0.2,
            horizontal_flip=True
        )
```

```
    else:
```

```
        data_gen = ImageDataGenerator(rescale=1./255)
```

```
    dataset = data_gen.flow_from_directory(
        data_dir,
        target_size=img_size,
        batch_size=batch_size,
        class_mode='categorical'
    )
```

```
    return dataset
```

```
train_data = load_data(os.path.join(data_dir), augment=True)
```

```
val_data = load_data(os.path.join(data_dir))
```

```
Found 301 images belonging to 3 classes.
```

```
Found 301 images belonging to 3 classes.
```

```
def visualize_images(dataset, num_images=9):
```

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

```
    images, labels = [], []
```

```
for batch_images, batch_labels in dataset:
    images.extend(batch_images)
    labels.extend(batch_labels)
    if len(images) >= num_images:
        break

images = np.array(images[:num_images])
labels = np.array(labels[:num_images])

for i in range(num_images):
    plt.subplot(3, 3, i + 1)
    plt.imshow(images[i])
    plt.title(list(dataset.class_indices.keys())
[ np.argmax(labels[i]) ])
    plt.axis('off')
plt.show()

visualize_images(train_data_example)
```

Matang



Busuk



Matang



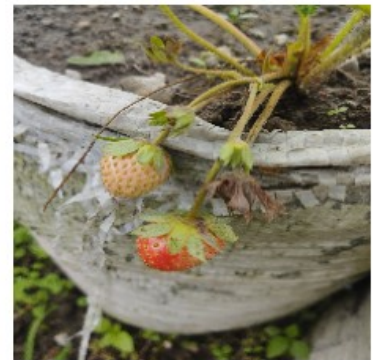
Matang



Matang



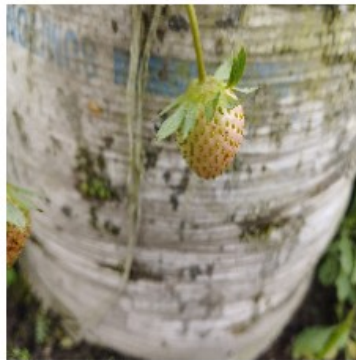
Mentah



Busuk



Mentah



Matang



```
def prepare_data(data_dir, img_size=(224, 224), batch_size=32):  
    train_dir = os.path.join(data_dir)  
    val_dir = os.path.join(data_dir)  
    test_dir =  
os.path.join('/Users/saktiyoga/Development/PMDPM/test_data')  
  
    train_data = load_data(train_dir, img_size, batch_size)  
    val_data = load_data(val_dir, img_size, batch_size)  
    test_data = load_data(test_dir, img_size, batch_size)  
  
    print(f"Train data size: {train_data.samples}")
```

```

print(f"Validation data size: {val_data.samples}")
print(f"Test data size: {test_data.samples}")

return train_data, val_data, test_data

data_dir = r'/Users/saktiyoga/Development/PMDPM/train_data'
train_data, val_data, test_data = prepare_data(data_dir)

Found 301 images belonging to 3 classes.
Found 301 images belonging to 3 classes.
Found 301 images belonging to 3 classes.
Train data size: 301
Validation data size: 301
Test data size: 301

def create_model(input_shape=(224, 224, 3), num_classes=3):
    base_model =
    tf.keras.applications.MobileNetV2(input_shape=input_shape,
                                      include_top=False,
                                      weights=None)

    model = tf.keras.Sequential([
        base_model,
        tf.keras.layers.GlobalAveragePooling2D(),
        tf.keras.layers.Dense(num_classes, activation='softmax')
    ])

    model.compile(optimizer='adam',
                  loss='categorical_crossentropy',
                  metrics=['accuracy'])

    model.summary()
    return model

model = create_model()
Model: "sequential_5"

```

Layer (type) Param #	Output Shape	
mobilenetv2_1.00_224 2,257,984 (Functional)	(None, 7, 7, 1280)	
global_average_pooling2d_2	(None, 1280)	

0	(GlobalAveragePooling2D)		

Total params: 2,261,827 (8.63 MB)

Trainable params: 2,227,715 (8.50 MB)

Non-trainable params: 34,112 (133.25 KB)

```
from tensorflow.keras.callbacks import EarlyStopping

def train_model(model, train_data, val_data, epochs=20):
    early_stopping = EarlyStopping(monitor='val_loss', patience=5,
    restore_best_weights=True)
    history = model.fit(train_data,
                        validation_data=val_data,
                        epochs=epochs,
                        callbacks=[early_stopping])

    return history

history = train_model(model, train_data, val_data, epochs=20)
```

```
Epoch 1/20
10/10 _____ 55s 6s/step - accuracy: 0.6752 - loss:
0.8182 - val_accuracy: 0.3355 - val_loss: 1.0988
Epoch 2/20
10/10 _____ 51s 5s/step - accuracy: 0.7641 - loss:
0.5618 - val_accuracy: 0.3322 - val_loss: 1.0987
Epoch 3/20
10/10 _____ 50s 5s/step - accuracy: 0.8241 - loss:
0.3878 - val_accuracy: 0.3322 - val_loss: 1.0987
Epoch 4/20
10/10 _____ 51s 5s/step - accuracy: 0.9352 - loss:
0.1967 - val_accuracy: 0.3322 - val_loss: 1.0996
Epoch 5/20
10/10 _____ 50s 5s/step - accuracy: 0.9479 - loss:
0.1539 - val_accuracy: 0.3322 - val_loss: 1.1026
Epoch 6/20
10/10 _____ 50s 5s/step - accuracy: 0.9825 - loss:
0.0700 - val_accuracy: 0.3322 - val_loss: 1.1044
Epoch 7/20
10/10 _____ 51s 5s/step - accuracy: 0.9610 - loss:
0.0807 - val_accuracy: 0.3322 - val_loss: 1.1163
Epoch 8/20
```

10/10 ————— 78s 8s/step - accuracy: 0.9251 - loss: 0.1287 - val\_accuracy: 0.3322 - val\_loss: 1.1185

```
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```

```
def evaluate_model(model, test_data):
    predictions = model.predict(test_data)
    y_pred = np.argmax(predictions, axis=1)
    y_true = test_data.classes

    cm = confusion_matrix(y_true, y_pred)
    disp = ConfusionMatrixDisplay(confusion_matrix=cm,

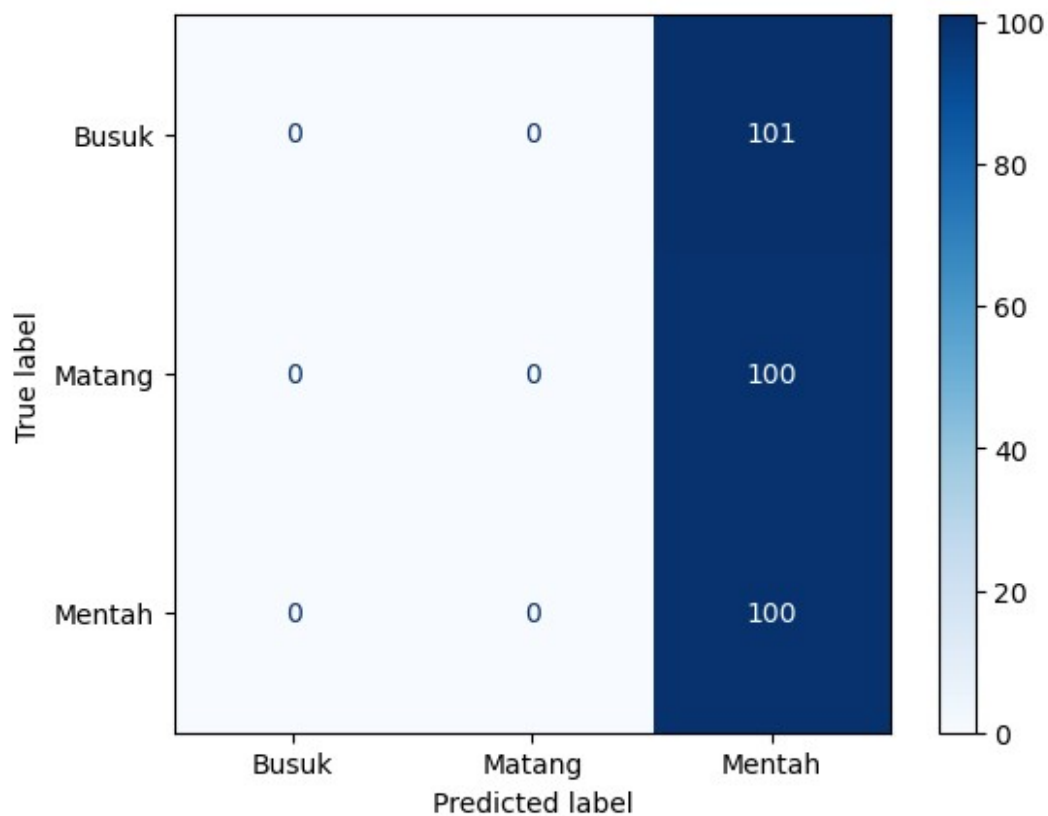
display_labels=list(test_data.class_indices.keys()))
    disp.plot(cmap=plt.cm.Blues)
    plt.show()

    test_data.reset()
    images, labels = next(test_data)
    for i in range(min(9, len(images))):
        plt.subplot(3, 3, i + 1)
        plt.imshow(images[i])
        pred_label = list(test_data.class_indices.keys())
[ np.argmax(predictions[i])]
        true_label = list(test_data.class_indices.keys())
[ np.argmax(labels[i])]
        plt.title(f"Pred: {pred_label}\nTrue: {true_label}")
        plt.axis('off')
    plt.show()
```

```
evaluate_model(model, test_data)
```

10/10 ————— 44s 4s/step





Pred: Mentah  
True: Matang



Pred: Mentah  
True: Busuk



Pred: Mentah  
True: Matang



Pred: Mentah  
True: Mentah



Pred: Mentah  
True: Mentah



Pred: Mentah  
True: Busuk



Pred: Mentah  
True: Busuk



Pred: Mentah  
True: Busuk



Pred: Mentah  
True: Matang





```
def plot_metrics(history):
    acc = history.history['accuracy']
    val_acc = history.history['val_accuracy']
    loss = history.history['loss']
    val_loss = history.history['val_loss']

    epochs_range = range(len(acc))

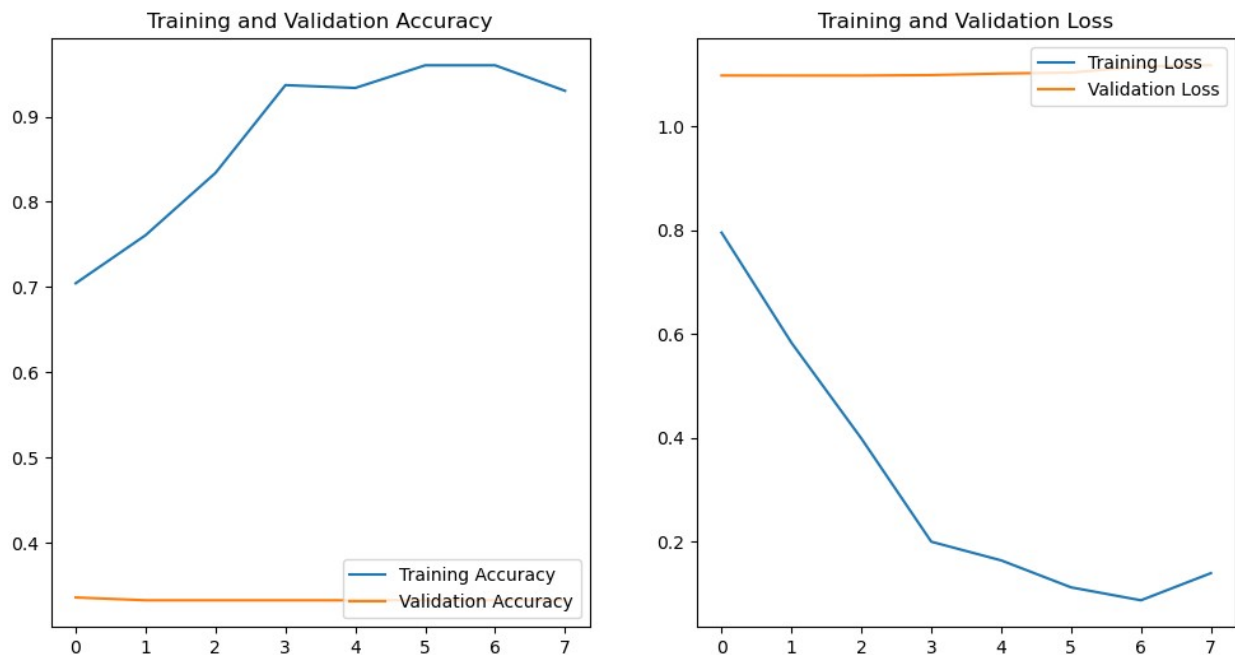
    plt.figure(figsize=(12, 6))

    plt.subplot(1, 2, 1)
    plt.plot(epochs_range, acc, label='Training Accuracy')
    plt.plot(epochs_range, val_acc, label='Validation Accuracy')
    plt.legend(loc='lower right')
    plt.title('Training and Validation Accuracy')

    plt.subplot(1, 2, 2)
    plt.plot(epochs_range, loss, label='Training Loss')
    plt.plot(epochs_range, val_loss, label='Validation Loss')
    plt.legend(loc='upper right')
    plt.title('Training and Validation Loss')

    plt.show()

plot_metrics(history)
```



```
def predict_strawberry(image_path, model):
```

```

img = tf.keras.preprocessing.image.load_img(
    image_path,
    target_size=(224, 224)
)
img_array = tf.keras.preprocessing.image.img_to_array(img)
img_array = tf.expand_dims(img_array, 0) / 255.0

predictions = model.predict(img_array)
predicted_class = categories[np.argmax(predictions[0])]

print(f'Predicted Strawberry Condition: {predicted_class}')
plt.imshow(plt.imread(image_path))
plt.title(f'Predicted: {predicted_class}')
plt.show()

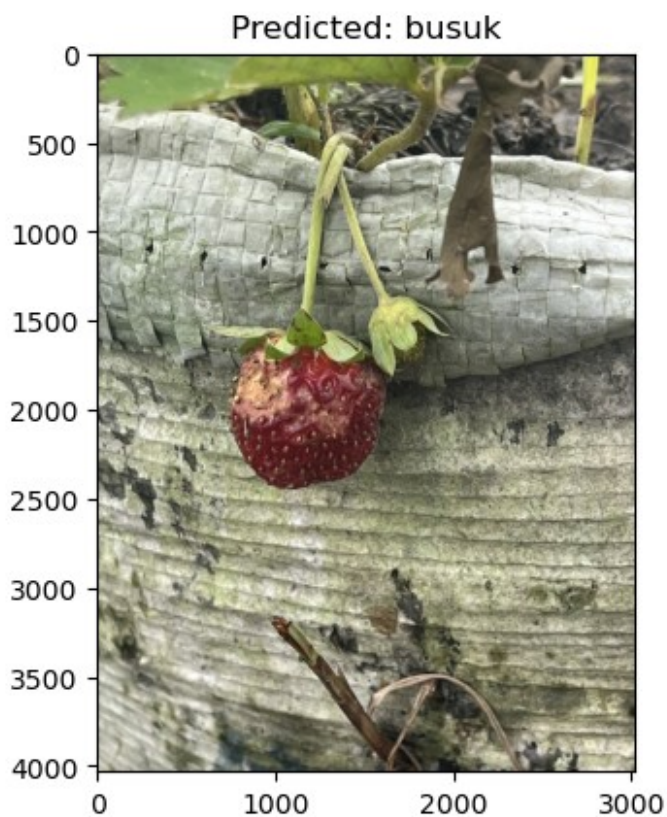
```

```

predict_strawberry('/Users/saktiyoga/Development/PMDPM/test_data/Busuk
/busuk_07.jpg', model)

```

1/1 ————— 0s 245ms/step  
Predicted Strawberry Condition: busuk



# VGG

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

import os
from PIL import Image

input_folder = r'C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER
5\ML\Tubes\train_data'
output_folder = r'C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER
5\ML\Tubes\train_data_konversi'

os.makedirs(output_folder, exist_ok=True)

def convert_images(input_folder, output_folder, target_format="JPEG"):
    for root, _, files in os.walk(input_folder):
        for filename in files:
            input_path = os.path.join(root, filename)
            relative_path = os.path.relpath(root, input_folder)
            output_subfolder = os.path.join(output_folder,
relative_path)
            os.makedirs(output_subfolder, exist_ok=True)

            try:
                with Image.open(input_path) as img:
                    if img.mode != "RGB":
                        img = img.convert("RGB")

                    new_filename = os.path.splitext(filename)[0] + f".
{target_format.lower()}"
                    output_path = os.path.join(output_subfolder,
new_filename)

                    img.save(output_path, target_format)
                    print(f"Berhasil mengonversi: {input_path} ->
{output_path}")
            except Exception as e:
                print(f"Error saat mengonversi {input_path}: {e}")

convert_images(input_folder, output_folder, target_format="JPEG")

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_01.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_01.jpeg
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_02.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_02.jpeg
```

[illegible]

[illegible]

[illegible]

Busuk\busuk\_38.jpeg

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_39.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_39.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_40.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Busuk\busuk_40.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_41.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 41.jpeg
```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_42.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 42.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_43.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 43.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_44.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 44.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_45.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 45.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_46.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 46.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_47.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 47.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_48.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 48.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_49.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 49.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_50.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 50.jpeg

```

[illegible]



[illegible]

[illegible]

Busuk\busuk\_86.jpeg

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_87.png -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_87.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_88.png -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_88.jpeg

```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_89.png -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_89.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_90.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Busuk\busuk_90.jpeg
```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_91.jpeg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 91.jpeg

```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_92.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 92.jpeg
```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_93.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 93.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_94.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk_94.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_95.png -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 95.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_96.png -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 96.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_97.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 97.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Busuk\busuk_98.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Busuk\busuk 98.jpeg

```

[illegible]

[illegible]

Matang\matang\_23.jpeg

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_24.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_24.jpeg

```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Matang\matang_25.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_25.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_26.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_26.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_27.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_27.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_28.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_28.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Matang\matang_29.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_29.jpeg
```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Matang\matang_30.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_30.jpeg
```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_31.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_31.jpeg

```

```
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_32.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_32.jpeg
```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_33.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang 33.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_34.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_34.jpeg

```

```

Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\
SEMESTER 5\ML\Tubes\train_data\Matang\matang_35.jpg -> C:\Users\vina
qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\
Matang\matang_35.jpeg

```



[illegible]



[illegible]



[illegible]

```
SEMESTER 5\ML\Tubes\train_data\Matang\matang_97.JPG -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_97.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Matang\matang_98.JPG -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_98.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Matang\matang_99.JPG -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Matang\matang_99.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_01.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_01.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_02.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_02.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_03.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_03.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_04.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_04.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_05.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_05.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_06.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_06.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_07.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_07.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_08.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_08.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data\Mentah\mentah_09.jpg -> C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi\Mentang\mentah_09.jpeg  
Berhasil mengonversi: C:\Users\vina qhuroutu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train data\Mentah\mentah 10.jpg -> C:\Users\vina
```

[illegible]





[illegible]







[illegible]

qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\_konversi\Mentah\mentah\_95.jpeg  
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\Mentah\mentah\_96.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\_konversi\Mentah\mentah\_96.jpeg  
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\Mentah\mentah\_97.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\_konversi\Mentah\mentah\_97.jpeg  
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\Mentah\mentah\_98.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\_konversi\Mentah\mentah\_98.jpeg  
Berhasil mengonversi: C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\Mentah\mentah\_99.jpg -> C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train\_data\_konversi\Mentah\mentah\_99.jpeg

```
data_dir = r"C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\train_data_konversi"
```

```
img_size = 180
```

```
batch = 32
```

```
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) * batch  
print("Total Images: ", total_count)
```

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

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

```
train_ds = dataset.take(train_count // batch)  
val_ds = dataset.skip(train_count // batch).take(val_count // batch)  
test_ds = dataset.skip(train_count // batch + val_count // batch).take(test_count // batch)
```

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

Found 301 files belonging to 3 classes.

Total Images: 320

Train Images: 256

Validation Images: 32

Test Images: 32

Class Names: ['Busuk', 'Matang', 'Mentah']

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')

Mentah



Busuk



Matang



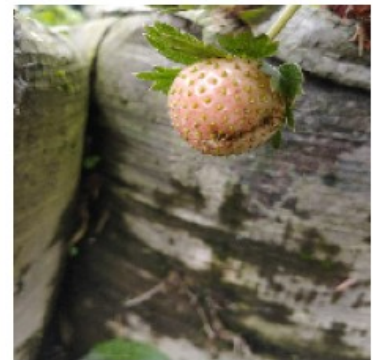
Mentah



Busuk



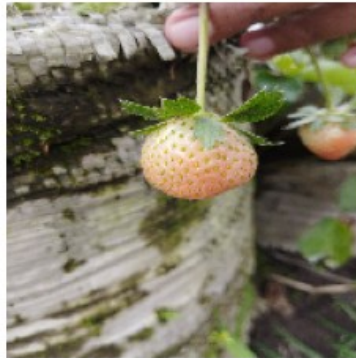
Mentah



Busuk



Mentah



Matang



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

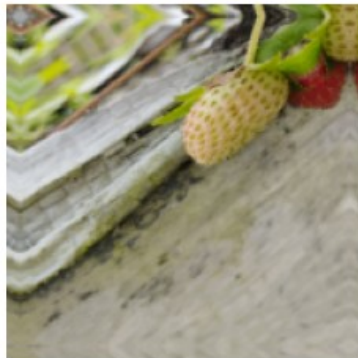
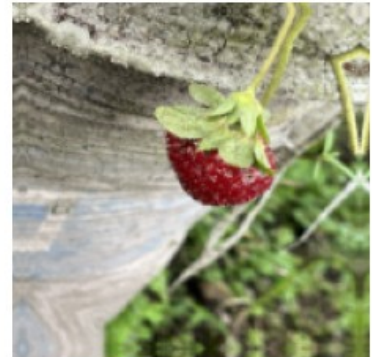
```
(32, 180, 180, 3)
```

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





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

def vgg16(input_shape, n_classes):
    base_model = tf.keras.applications.VGG16(weights='imagenet',
include_top=False, input_shape=input_shape)
    base_model.trainable = False

    model = models.Sequential()
    model.add(base_model)
    model.add(layers.Flatten())
    model.add(layers.Dense(256, activation='relu'))
```



```

model.add(layers.BatchNormalization())
model.add(layers.Dropout(0.5))
model.add(layers.Dense(n_classes, activation='softmax'))

# Compile the model

model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=0.0001)
,
               loss='sparse_categorical_crossentropy',
               metrics=['accuracy'])

return model

input_shape = (180, 180, 3)
n_classes = len(class_names) # 3 classes: ['Busuk', 'Matang',
'Mentah']

tf.keras.backend.clear_session()

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

Model: "sequential"

```

Layer (type) Param #	Output Shape	
vgg16 (Functional) 14,714,688	(None, 5, 5, 512)	
flatten (Flatten) 0	(None, 12800)	
dense (Dense) 3,277,056	(None, 256)	
batch_normalization (BatchNormalization) 1,024	(None, 256)	
dropout (Dropout) 0	(None, 256)	

dense_1 (Dense)	(None, 3)	
771		

Total params: 17,993,539 (68.64 MB)

Trainable params: 3,278,339 (12.51 MB)

Non-trainable params: 14,715,200 (56.13 MB)

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

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

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

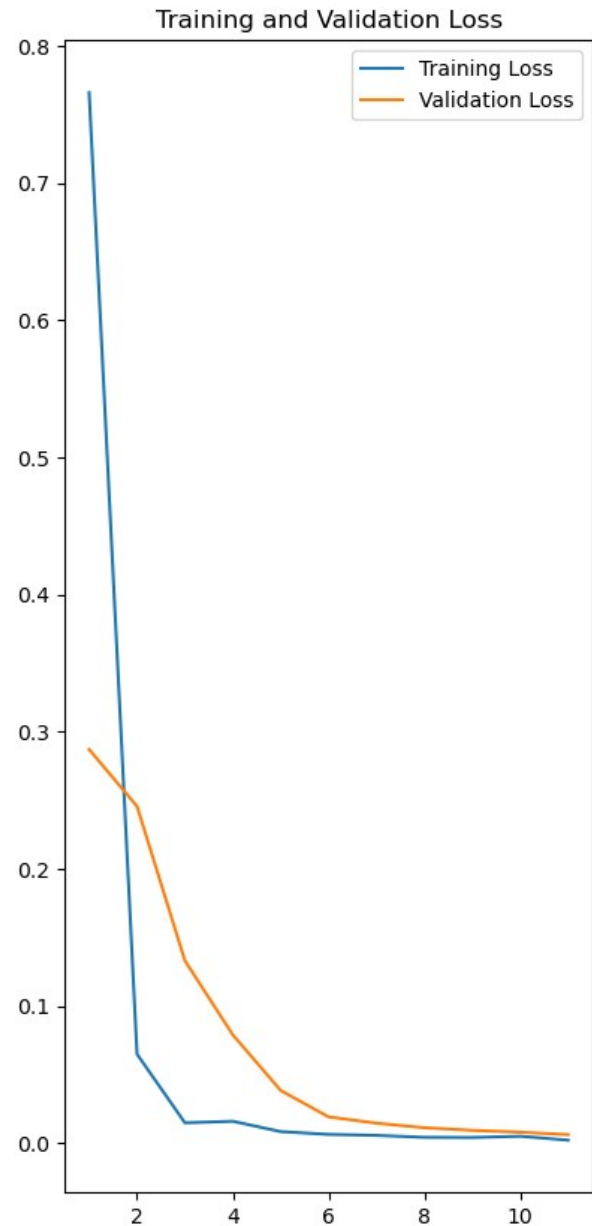
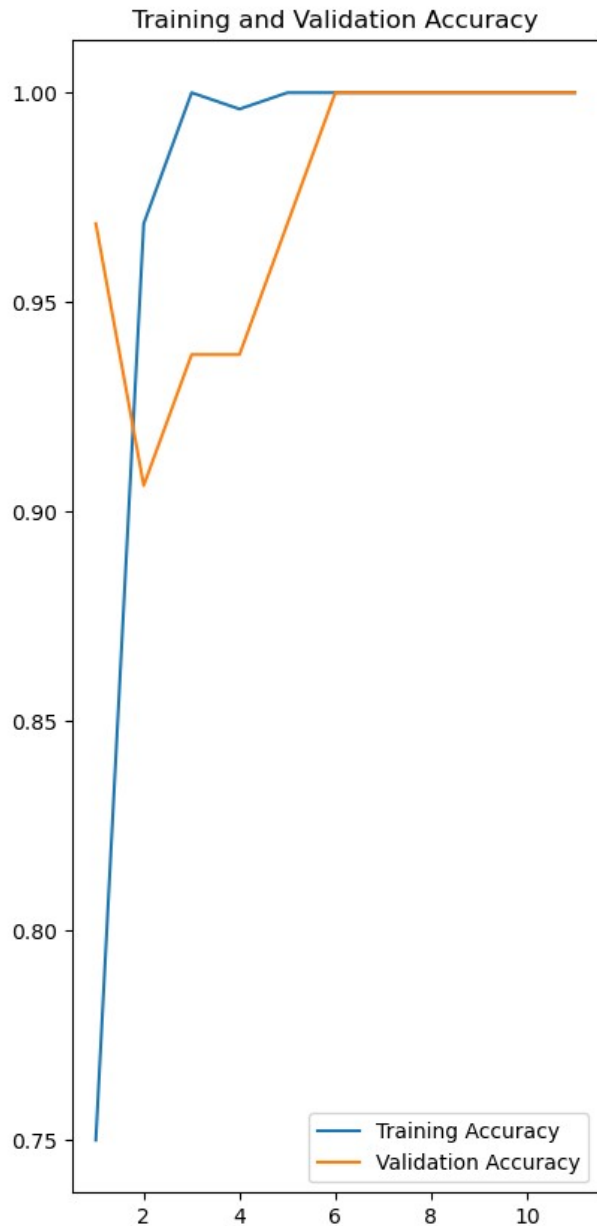
history = model.fit(train_ds,
                    epochs=30,
                    validation_data=val_ds,
                    callbacks=[early_stopping])

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.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.title('Training and Validation Loss')
plt.show()

Epoch 1/30
8/8 _____ 10s 1s/step - accuracy: 0.6639 - loss: 1.0554
- val_accuracy: 0.9688 - val_loss: 0.2871
```

```
Epoch 2/30
8/8 _____ 8s 1s/step - accuracy: 0.9583 - loss: 0.0799
- val_accuracy: 0.9062 - val_loss: 0.2457
Epoch 3/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0220
- val_accuracy: 0.9375 - val_loss: 0.1331
Epoch 4/30
8/8 _____ 9s 1s/step - accuracy: 0.9965 - loss: 0.0141
- val_accuracy: 0.9375 - val_loss: 0.0790
Epoch 5/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0092
- val_accuracy: 0.9688 - val_loss: 0.0385
Epoch 6/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0051
- val_accuracy: 1.0000 - val_loss: 0.0192
Epoch 7/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0076
- val_accuracy: 1.0000 - val_loss: 0.0147
Epoch 8/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0057
- val_accuracy: 1.0000 - val_loss: 0.0113
Epoch 9/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0036
- val_accuracy: 1.0000 - val_loss: 0.0095
Epoch 10/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0038
- val_accuracy: 1.0000 - val_loss: 0.0081
Epoch 11/30
8/8 _____ 9s 1s/step - accuracy: 1.0000 - loss: 0.0017
- val_accuracy: 1.0000 - val_loss: 0.0063
```



```
model.save('VGG CNN_Matplotlib.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
from tensorflow.keras.models import load_model
```

```
model = load_model(r'C:\Users\vina qhurotu aini\Documents\KULIAH\SEMESTER 5\ML\Tubes\VGG CNN_Matplotlib.h5') # Ganti dengan path model
```

Anda

```
class_names = ['Busuk', 'Matang', 'Mentah']
```

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.

```
def classify_images(image_path, save_path='predicted_image.jpg'):
    try:
        input_image = tf.keras.utils.load_img(image_path,
        target_size=(180, 180)) # Sesuaikan ukuran jika perlu
        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

        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.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\vina ghurotu aini\Documents\
KULIAH\SEMESTER 5\ML\Tubes\test_data\Busuk\busuk_09.jpg',
save_path='busuk9.jpg')
print(result)
```

1/1 ————— 0s 213ms/step

Prediksi: Busuk

Confidence: 57.55%

Prediksi: Busuk dengan confidence 57.55%. Gambar asli disimpan di busuk9.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)
    )

Found 29 files belonging to 3 classes.

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

1/1 ————— 1s 1s/step

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)

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

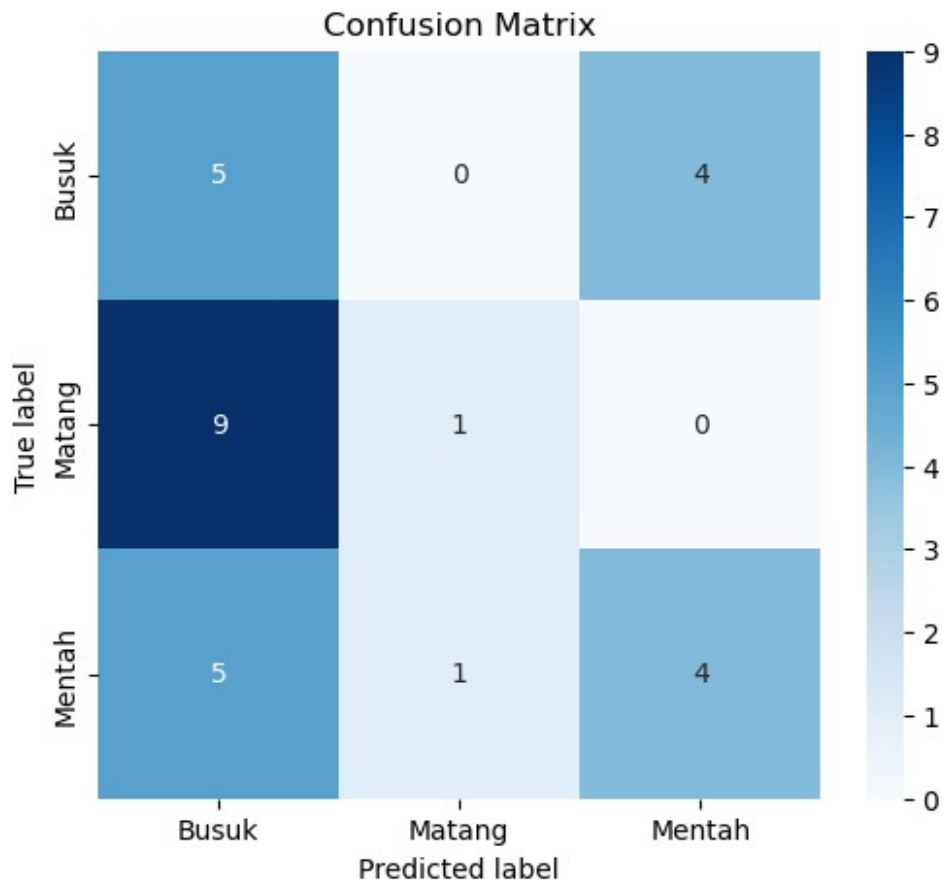
accuracy = tf.reduce_sum(tf.linalg.diag_part(conf_mat)) /
tf.reduce_sum(conf_mat)

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)

# Visualisasi Confusion Matrix
plt.figure(figsize=(6, 5))
sns.heatmap(conf_mat.numpy(), annot=True, fmt='d', cmap='Blues',
            xticklabels=["Busuk", "Matang", "Mentah"],
            yticklabels=["Busuk", "Matang", "Mentah"])
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())
```

Confusion Matrix:

```
[[5 0 4]
 [9 1 0]
 [5 1 4]]
```

Akurasi: 0.3448275862068966

Presisi: [0.26315789 0.5 0.5 ]

Recall: [0.55555556 0.1 0.4 ]

F1 Score: [0.35714286 0.16666667 0.44444444]

# ALEX NET

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

# Load data
data_dir = r"C:\Users\Sherlyna Alfelia\Documents\KULIAH\SMT 5\ML\
TUBES_UAS\train_data"
data = tf.keras.utils.image_dataset_from_directory(data_dir, seed=123,
image_size=(180, 180))

print(data.class_names)

class_names = data.class_names # ['Busuk', 'Matang', 'Mentah']

Found 301 files belonging to 3 classes.
['Busuk', 'Matang', 'Mentah']

data_dir = r"C:\Users\Sherlyna Alfelia\Documents\KULIAH\SMT 5\ML\
TUBES_UAS\train_data"
img_size = 180
batch = 32
validation_split = 0.1

# Membagi dataset menjadi train dan validation
train_ds = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
    validation_split=validation_split,
    subset="training"
)

val_ds = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    seed=123,
    image_size=(img_size, img_size),
    batch_size=batch,
    validation_split=validation_split,
    subset="validation"
)

# Cek jumlah gambar
print("Total Images: ", len(train_ds) * batch + len(val_ds) * batch)
print("Train Images: ", len(train_ds) * batch)
print("Validation Images: ", len(val_ds) * batch)

Found 301 files belonging to 3 classes.
Using 271 files for training.
```



Found 301 files belonging to 3 classes.

Using 30 files for validation.

Total Images: 320

Train Images: 288

Validation Images: 32

*# Visualisasi data*

```
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')
```

```
plt.show()
```

Mentah



Busuk



Matang



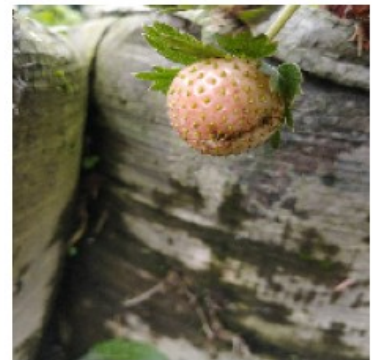
Mentah



Busuk



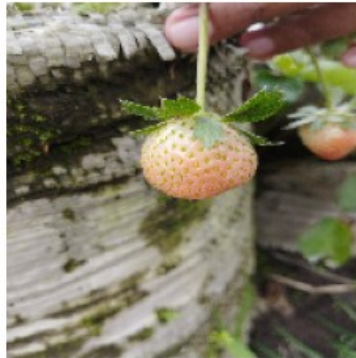
Mentah



Busuk



Mentah



Matang



```
# Preprocessing dataset
```

```
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
```

```
data_augmentation = models.Sequential([
```

```
    layers.RandomFlip("horizontal", input_shape=(img_size, img_size,
```

```
3)),
```

```
    layers.RandomRotation(0.1),
```

```
    layers.RandomZoom(0.1)
```

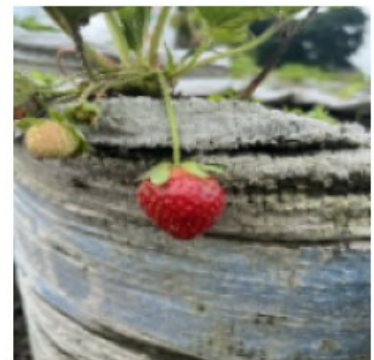
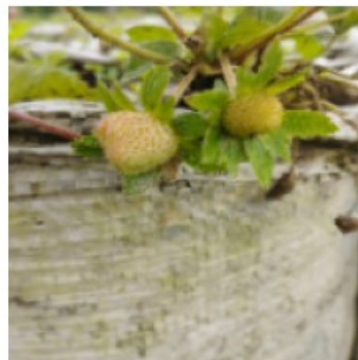
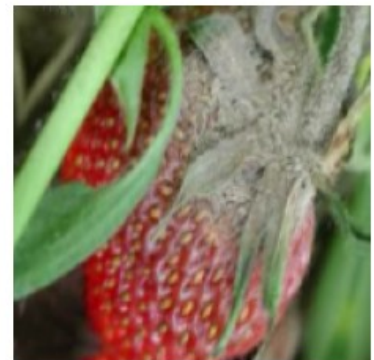
```

])

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

c:\anac\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)

```



```
from tensorflow.keras import layers, models

# Membuat model AlexNet dengan Batch Normalization
def alexnet(input_shape, n_classes):
    model = models.Sequential()

    # Layer 1
    model.add(layers.Conv2D(96, (11, 11), strides=(4, 4),
        activation='relu', input_shape=input_shape))
    model.add(layers.BatchNormalization())
    model.add(layers.MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
```



```

# Layer 2
model.add(layers.Conv2D(256, (5, 5), padding='same',
activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))

# Layer 3
model.add(layers.Conv2D(384, (3, 3), padding='same',
activation='relu'))
model.add(layers.BatchNormalization())

# Layer 4
model.add(layers.Conv2D(384, (3, 3), padding='same',
activation='relu'))
model.add(layers.BatchNormalization())

# Layer 5
model.add(layers.Conv2D(256, (3, 3), padding='same',
activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))

# Flatten dan Dense Layers
model.add(layers.Flatten())
model.add(layers.Dense(4096, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(0.01)))
model.add(layers.Dropout(0.5))
model.add(layers.Dense(4096, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(0.01)))
model.add(layers.Dropout(0.5))
model.add(layers.Dense(n_classes, activation='softmax'))

return model

```

```

# Pastikan input shape dan jumlah kelas sesuai
input_shape = (180, 180, 3)
n_classes = len(class_names)

```

```

# Clear Cache Keras menggunakan clear session
tf.keras.backend.clear_session()

```

```

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

```

```

c:\anac\Lib\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) 34,944	(None, 43, 43, 96)	
batch_normalization (BatchNormalization) 384	(None, 43, 43, 96)	
max_pooling2d (MaxPooling2D) 0	(None, 21, 21, 96)	
conv2d_1 (Conv2D) 614,656	(None, 21, 21, 256)	
batch_normalization_1 (BatchNormalization) 1,024	(None, 21, 21, 256)	
max_pooling2d_1 (MaxPooling2D) 0	(None, 10, 10, 256)	
conv2d_2 (Conv2D) 885,120	(None, 10, 10, 384)	
batch_normalization_2 (BatchNormalization) 1,536	(None, 10, 10, 384)	
conv2d_3 (Conv2D)	(None, 10, 10, 384)	

1,327,488		
batch_normalization_3	(None, 10, 10, 384)	
1,536	(BatchNormalization)	
conv2d_4 (Conv2D)	(None, 10, 10, 256)	
884,992		
batch_normalization_4	(None, 10, 10, 256)	
1,024	(BatchNormalization)	
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 256)	
0		
flatten (Flatten)	(None, 4096)	
0		
dense (Dense)	(None, 4096)	
16,781,312		
dropout (Dropout)	(None, 4096)	
0		
dense_1 (Dense)	(None, 4096)	
16,781,312		
dropout_1 (Dropout)	(None, 4096)	
0		
dense_2 (Dense)	(None, 3)	
12,291		

Total params: 37,327,619 (142.39 MB)

Trainable params: 37,324,867 (142.38 MB)

Non-trainable params: 2,752 (10.75 KB)

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

# Compile dengan optimizer Adam
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

9/9 ————— 23s 1s/step - accuracy: 0.4080 - loss: 109.9298

Epoch 2/30

c:\anac\Lib\contextlib.py:158: UserWarning: Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps\_per\_epoch \* epochs` batches. You may need to use the `.repeat()` function when building your dataset.

self.gen.throw(value)

c:\anac\Lib\site-packages\keras\src\callbacks\early\_stopping.py:155: UserWarning: Early stopping conditioned on metric `val\_accuracy` which is not available. Available metrics are: accuracy,loss

current = self.get\_monitor\_value(logs)

9/9 ————— 8s 894ms/step - accuracy: 0.4332 - loss: 78.5367

Epoch 3/30

9/9 ————— 8s 872ms/step - accuracy: 0.5318 - loss: 64.9256

Epoch 4/30

9/9 ————— 7s 824ms/step - accuracy: 0.5239 - loss: 58.5072

Epoch 5/30

9/9 ————— 8s 842ms/step - accuracy: 0.6050 - loss: 51.3457

Epoch 6/30



9/9 ————— 8s 855ms/step - accuracy: 0.6548 - loss: 45.0703  
Epoch 7/30  
9/9 ————— 8s 906ms/step - accuracy: 0.6395 - loss: 41.0658  
Epoch 8/30  
9/9 ————— 9s 963ms/step - accuracy: 0.6824 - loss: 36.7263  
Epoch 9/30  
9/9 ————— 8s 855ms/step - accuracy: 0.6978 - loss: 33.6807  
Epoch 10/30  
9/9 ————— 8s 864ms/step - accuracy: 0.7614 - loss: 28.9888  
Epoch 11/30  
9/9 ————— 8s 847ms/step - accuracy: 0.7177 - loss: 25.9854  
Epoch 12/30  
9/9 ————— 9s 960ms/step - accuracy: 0.7938 - loss: 23.6791  
Epoch 13/30  
9/9 ————— 8s 844ms/step - accuracy: 0.7513 - loss: 21.5451  
Epoch 14/30  
9/9 ————— 7s 820ms/step - accuracy: 0.7914 - loss: 19.2834  
Epoch 15/30  
9/9 ————— 9s 967ms/step - accuracy: 0.8574 - loss: 17.9925  
Epoch 16/30  
9/9 ————— 8s 855ms/step - accuracy: 0.8658 - loss: 15.8844  
Epoch 17/30  
9/9 ————— 8s 877ms/step - accuracy: 0.8551 - loss: 14.8671  
Epoch 18/30  
9/9 ————— 8s 853ms/step - accuracy: 0.8893 - loss: 14.3423  
Epoch 19/30  
9/9 ————— 8s 904ms/step - accuracy: 0.7516 - loss: 14.3587  
Epoch 20/30  
9/9 ————— 8s 833ms/step - accuracy: 0.8515 - loss: 12.0668  
Epoch 21/30  
9/9 ————— 7s 828ms/step - accuracy: 0.8450 - loss: 11.2583  
Epoch 22/30  
9/9 ————— 7s 845ms/step - accuracy: 0.9205 - loss:

```
10.2116
Epoch 23/30
9/9 _____ 8s 841ms/step - accuracy: 0.9442 - loss:
9.4293
Epoch 24/30
9/9 _____ 8s 869ms/step - accuracy: 0.9140 - loss:
8.8839
Epoch 25/30
9/9 _____ 8s 864ms/step - accuracy: 0.9533 - loss:
8.0947
Epoch 26/30
9/9 _____ 8s 891ms/step - accuracy: 0.9307 - loss:
8.2085
Epoch 27/30
9/9 _____ 8s 869ms/step - accuracy: 0.9046 - loss:
7.5800
Epoch 28/30
9/9 _____ 7s 850ms/step - accuracy: 0.9596 - loss:
6.8011
Epoch 29/30
9/9 _____ 7s 831ms/step - accuracy: 0.9357 - loss:
6.5429
Epoch 30/30
9/9 _____ 8s 839ms/step - accuracy: 0.9490 - loss:
6.1547
```

```
# Cek kunci dalam history
```

```
print(history.history.keys())
```

```
# Rentang epoch
```

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

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

```
# Subplot untuk akurasi
```

```
plt.subplot(1, 2, 1)
```

```
plt.plot(epochs_range, history.history['accuracy'], label='Training Accuracy')
```

```
# Sesuaikan berdasarkan kunci yang ditemukan
```

```
if 'val_accuracy' in history.history:
```

```
    plt.plot(epochs_range, history.history['val_accuracy'],  
            label='Validation Accuracy')
```

```
elif 'val_acc' in history.history:
```

```
    plt.plot(epochs_range, history.history['val_acc'],  
            label='Validation Accuracy')
```

```
plt.legend(loc='lower right')
```

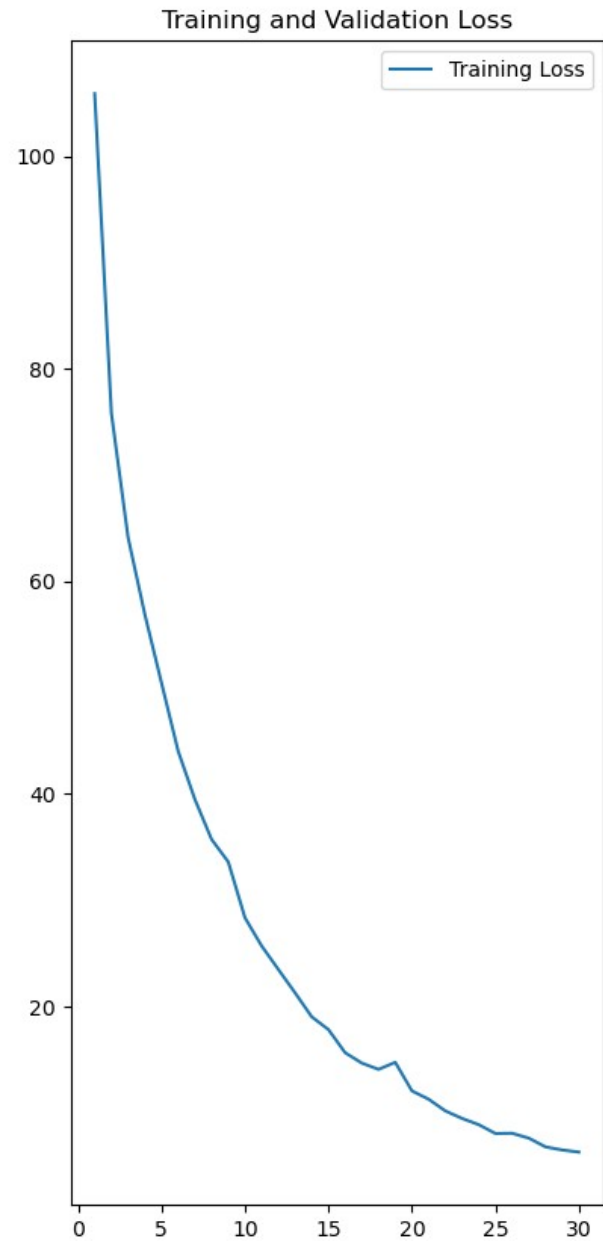
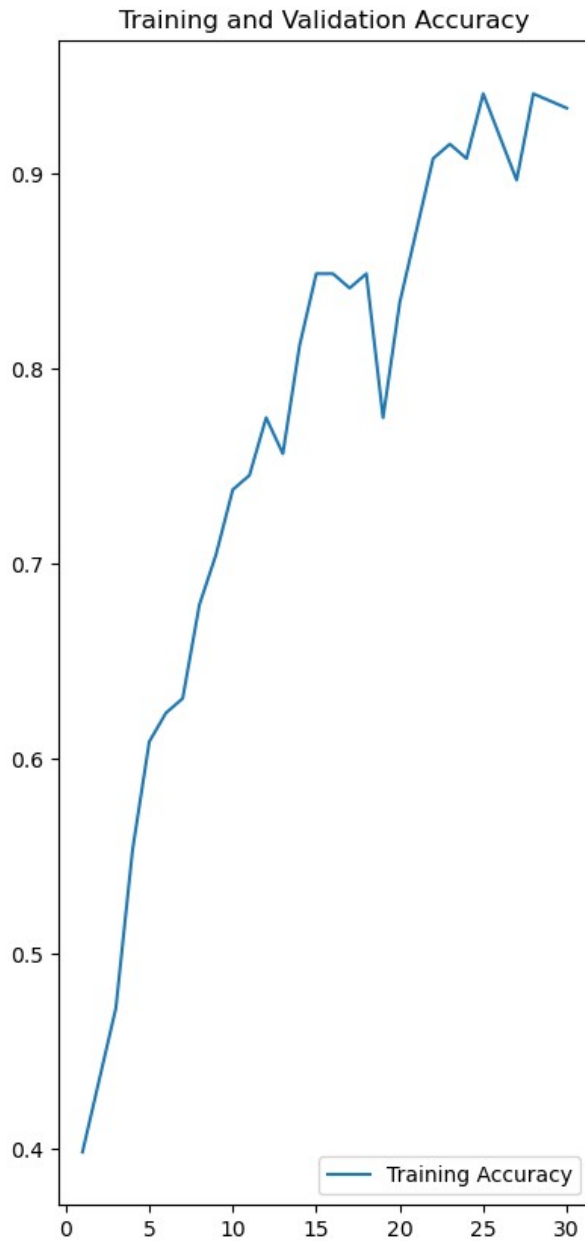
```
plt.title('Training and Validation Accuracy')
```

```
# Subplot untuk loss
plt.subplot(1, 2, 2)
plt.plot(epochs_range, history.history['loss'], label='Training Loss')

# Sesuaikan untuk loss validasi
if 'val_loss' in history.history:
    plt.plot(epochs_range, history.history['val_loss'],
label='Validation Loss')
elif 'val_loss' in history.history:
    plt.plot(epochs_range, history.history['val_loss'],
label='Validation Loss')

plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()

dict_keys(['accuracy', 'loss'])
```



```
model.save('BestModel_AlexNet_Matplotlib.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')`.

```
from tensorflow.keras.models import load_model
from PIL import Image
```

```
# Load the trained model
```

```

model = load_model(r'C:\Users\Sherlyna Alfelia\Documents\KULIAH\SMT 5\
ML\TUBES_UAS\BestModel_AlexNet_Matplotlib.h5') # Ganti dengan path
model Anda
class_names = ['Busuk', 'Matang', 'Mentah']

# 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
result = classify_images(r'C:\Users\Sherlyna Alfelia\Documents\KULIAH\
SMT 5\ML\TUBES_UAS\test_data\Matang\matang_06.jpg',
save_path='matang_06.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 480ms/step

Prediksi: Matang

Confidence: 57.61%

Prediksi: Matang dengan confidence 57.61%. Gambar asli disimpan di matang\_06.jpg.

```

import seaborn as sns

# 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
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)

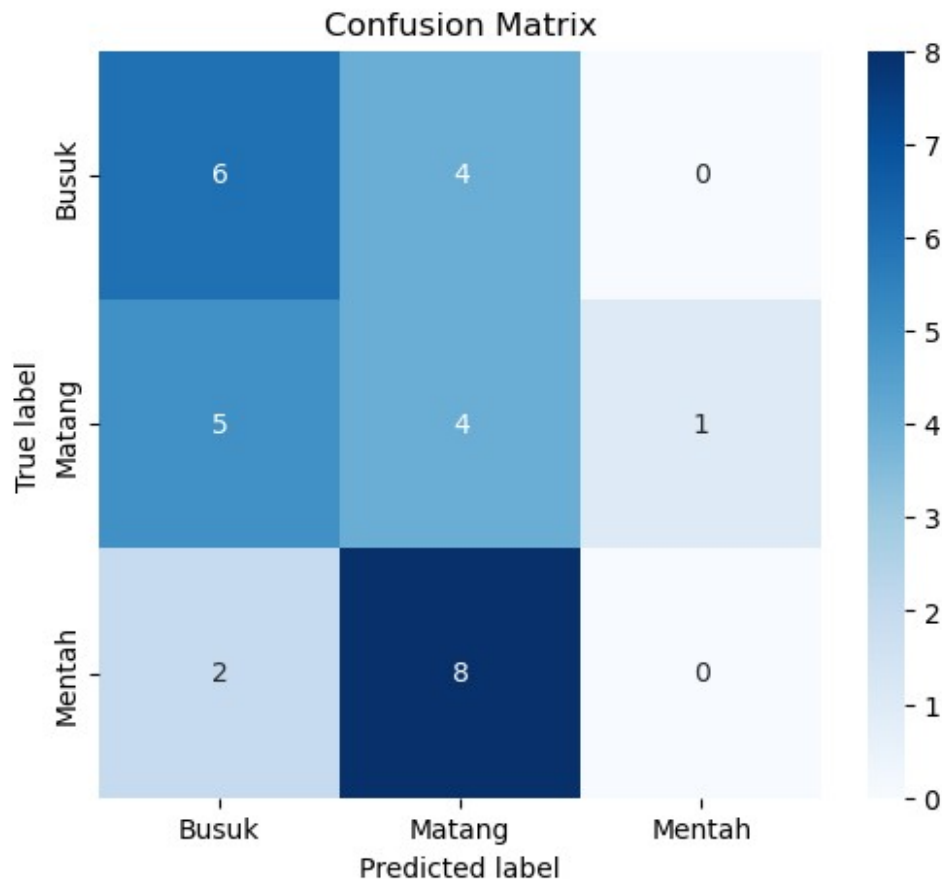
# Menghitung F1 Score
f1_score = 2 * (precision * recall) / (precision + recall)

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

```

```
# Menampilkan hasil
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 955ms/step



```
Confusion Matrix:
[[6 4 0]
 [5 4 1]
 [2 8 0]]
Akurasi: 0.3333333333333333
Presisi: [0.46153846 0.25 0. ]
Recall: [0.6 0.4 0. ]
F1 Score: [0.52173913 0.30769231 nan]
```

```

import streamlit as st
import tensorflow as tf
import numpy as np
from tensorflow.keras.models import load_model
from PIL import Image

# Load the pre-trained model
model = load_model(r'D:\coolyeah\semester5\ml\tubes_uas\gNet5.h5') # Adjust the
path to your model
class_names = ['Busuk', 'Matang', 'Mentah']

# Function to preprocess and classify image
def classify_image(image):
    try:
        # Preprocess the image
        input_image = image.resize((180, 180)) # Resize to match model input
        input_image_array = np.array(input_image) # Convert to numpy array
        input_image_exp_dim = np.expand_dims(input_image_array, axis=0) # Add
batch dimension

        # Predict using the model
        predictions = model.predict(input_image_exp_dim)
        result = tf.nn.softmax(predictions[0]) # Apply softmax for probability

        # Get class with highest confidence
        class_idx = np.argmax(result)
        confidence_scores = result.numpy()
        return class_names[class_idx], confidence_scores
    except Exception as e:
        return "Error", str(e)

# Function to create a custom progress bar
def custom_progress_bar(confidence, color1, color2):
    percentage1 = confidence[0] * 100 # Confidence for class 0 (Busuk)
    percentage2 = confidence[1] * 100 # Confidence for class 1 (Matang)
    percentage3 = confidence[2] * 100 # Confidence for class 2 (Mentah)
    progress_html = f"""
<div style="border: 1px solid #ddd; border-radius: 5px; overflow: hidden;
width: 100%; font-size: 14px;">
    <div style="width: {percentage1:.2f}%; background: #FF4136; color: white;
text-align: center; height: 24px; float: left;">
        {percentage1:.2f}% Busuk
    </div>
    <div style="width: {percentage2:.2f}%; background: #007BFF; color: white;
text-align: center; height: 24px; float: left;">
        {percentage2:.2f}% Matang
    </div>
    <div style="width: {percentage3:.2f}%; background: #2ECC40; color: white;
text-align: center; height: 24px; float: left;">
        {percentage3:.2f}% Mentah
    </div>
</div>
"""
    st.sidebar.markdown(progress_html, unsafe_allow_html=True)

# Streamlit UI
st.title("Prediksi Strawberry") # 4 digit npm terakhir

# Upload multiple files in the main page

```



```

uploaded_files = st.file_uploader("Unggah Gambar (Beberapa diperbolehkan)",
type=["jpg", "png", "jpeg"], accept_multiple_files=True)

# Sidebar for prediction button and results
if st.sidebar.button("Prediksi"):
    if uploaded_files:
        st.sidebar.write("### Hasil Prediksi")
        for uploaded_file in uploaded_files:
            image = Image.open(uploaded_file) # Open the uploaded image

            # Perform prediction
            label, confidence = classify_image(image)

            if label != "Error":
                # Display prediction results
                st.sidebar.write(f"**Nama File:** {uploaded_file.name}")
                st.sidebar.markdown(f"<h4 style='color: #007BFF;'>Prediksi: {label}
</h4>", unsafe_allow_html=True)

                # Display confidence scores
                st.sidebar.write("**Confidence:**")
                for i, class_name in enumerate(class_names):
                    st.sidebar.write(f"- {class_name}: {confidence[i] * 100:.2f}%")

                # Display custom progress bar
                custom_progress_bar(confidence, "#FF4136", "#007BFF")

                st.sidebar.write("---")
            else:
                st.sidebar.error(f"Kesalahan saat memproses gambar
{uploaded_file.name}: {confidence}")
        else:
            st.sidebar.error("Silakan unggah setidaknya satu gambar untuk diprediksi.")

# Preview images in the main page
if uploaded_files:
    st.write("### Preview Gambar")
    for uploaded_file in uploaded_files:
        image = Image.open(uploaded_file)
        st.image(image, caption=f"{uploaded_file.name}", use_column_width=True)

```