classification-using-resnet-50-v2

March 28, 2024

[35]: trainpath=r"D:\Winter Semester 3\AIML\Project\Alzheiemer_image\Alzheimer_s_

```
⇔Dataset\train"
      testpath=r"D:\Winter Semester 3\AIML\Project\Alzheiemer_image\Alzheimer_s_
       ⇔Dataset\test"
[36]: import tensorflow
      from tensorflow.keras.layers import Dense, Flatten, Input, Dropout
      from tensorflow.keras.models import Model
      from tensorflow.keras.preprocessing import image
      from tensorflow.keras.preprocessing.image import ImageDataGenerator as IDG, __
       →load_img
      from tensorflow.keras.applications import Xception
      from tensorflow.keras import layers,models
      from tensorflow.keras.models import Sequential
      from tensorflow.keras.layers import
       SeparableConv2D, BatchNormalization, GlobalAveragePooling2D
      from tensorflow.keras.models import load_model
      from PIL import Image
      from tensorflow.keras.applications.xception import preprocess_input
      import numpy as np
[37]: import tensorflow
      from tensorflow.keras.applications import ResNet50
      from tensorflow.keras.layers import Dense, Flatten
      from tensorflow.keras.models import Model
      from tensorflow.keras.preprocessing.image import ImageDataGenerator
      img_size=180
      img_size_dim=[180,180]
      dimension=(img_size,img_size)
      zoom=[0.99,1.01]
      bright=[0.8,1.2]
      fill_mode='constant'
      data_format='channels_last'
      dir="D:/Winter Semester 3/AIML/Project\Alzheiemer_image/Alzheimer_s Dataset/
       ⇔train"
      data_gen = ImageDataGenerator(rescale=1./
       -255, brightness_range=bright,zoom_range=zoom,data_format=data_format,fill_mode=fill_mode,hor
```

```
#data_gen=IDG(rescale=1./
       $\to 255, brightness range=bright, zoom range=zoom, data format=data format, fill mode=fill mode, hor
      train_data_gen=data_gen.flow_from_directory(directory="D:/Winter Semester 3/
       →AIML/Project/Alzheiemer image/Alzheimer s Dataset/
       strain",target_size=dimension,batch_size=6000,shuffle=False)
     Found 5121 images belonging to 4 classes.
[38]: # Fetch a batch of training data and labels
      train_data, train_labels = train_data_gen.__next__()
      # Print the shape of the training data and labels
      print(train_data.shape, train_labels.shape)
     (5121, 180, 180, 3) (5121, 4)
[39]: #handling imbalanced classes
      from imblearn.over_sampling import SMOTE
      sm=SMOTE(random_state=47)
      train_data,train_labels=sm.fit_resample(train_data.
       →reshape(-1,img_size*img_size*3),train_labels)
      train_data=train_data.reshape(-1,img_size,img_size,3)
      print(train_data.shape,train_labels.shape)
     (10240, 180, 180, 3) (10240, 4)
[40]: from sklearn.model selection import train test split
      train_data,test_data,train_labels,test_labels =_u
       otrain_test_split(train_data,train_labels,test_size=0.2,random_state=47)
      train_data, val_data, train_labels, val_labels =_
       otrain_test_split(train_data,train_labels,test_size=0.2,random_state=47)
[41]: import tensorflow as tf
      from tensorflow.keras.applications import ResNet50
      from tensorflow.keras.layers import Dense, Flatten
      from tensorflow.keras.models import Model
      from tensorflow.keras.preprocessing.image import ImageDataGenerator
      # Define input shape and number of classes
      input\_shape = (224, 224, 3)
      num_classes = 4  # Alzheimer's and non-Alzheimer's
      # Load ResNet-50 model with pre-trained weights
      base_model = ResNet50(weights='imagenet', include_top=False,_
       →input_shape=input_shape)
      # Freeze some layers and fine-tune others
      for layer in base_model.layers[:-10]:
```

[57]: model.summary()

Model: "sequential_3"

Layer (type)	Output	Shape	Param #
conv2d_4 (Conv2D)	(None,	178, 178, 32)	896
<pre>max_pooling2d_4 (MaxPooling2D)</pre>	(None,	89, 89, 32)	0
conv2d_5 (Conv2D)	(None,	87, 87, 64)	18,496
<pre>max_pooling2d_5 (MaxPooling2D)</pre>	(None,	43, 43, 64)	0
flatten_4 (Flatten)	(None,	118336)	0
dense_19 (Dense)	(None,	64)	7,573,568
dense_20 (Dense)	(None,	4)	260

Total params: 7,593,220 (28.97 MB)

Trainable params: 7,593,220 (28.97 MB)

Non-trainable params: 0 (0.00 B)

```
[46]: mt=[tensorflow.keras.metrics.CategoricalAccuracy(name='Accuracy'),tensorflow.
       ⇔keras.metrics.AUC(name='AUC')]
 [1]: import tensorflow as tf
      from tensorflow.keras.applications import ResNet50
      from tensorflow.keras.layers import Dense, Flatten
      from tensorflow.keras.models import Model
      from tensorflow.keras.preprocessing.image import ImageDataGenerator
      # Define input shape and number of classes
      input shape = (224, 224, 3)
      num_classes = 4  # Alzheimer's and non-Alzheimer's
      # Load ResNet-50 model with pre-trained weights
      base_model = ResNet50(weights='imagenet', include_top=False,_
       →input_shape=input_shape)
      # Freeze some layers and fine-tune others
      for layer in base_model.layers[:-10]:
          layer.trainable = False
      # Add custom classification layers on top of ResNet-50
      x = Flatten()(base_model.output)
      x = Dense(256, activation='relu')(x) # Increase the number of neurons
      x = Dense(128, activation='relu')(x)
      predictions = Dense(num_classes, activation='softmax')(x)
      # Create the final model
      model = Model(inputs=base_model.input, outputs=predictions)
      # Compile the model with a lower learning rate
      model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=0.0001),
                    loss='sparse_categorical_crossentropy', # for integer labels
                    metrics=['accuracy'])
      # Define data generators with data augmentation
      train_datagen = ImageDataGenerator(
          rescale=1./255,
          validation_split=0.2,
          rotation_range=20,
          width_shift_range=0.2,
          height_shift_range=0.2,
          shear range=0.2,
          zoom_range=0.2,
          horizontal flip=True
      train_generator = train_datagen.flow_from_directory(
```

```
"D:/Winter Semester 3/AIML/Project/Alzheiemer_image/Alzheimer_s Dataset/
  target_size=(224, 224),
    batch size=32,
    class_mode='binary', # binary classification
    subset='training'
validation_generator = train_datagen.flow_from_directory(
    r"D:/Winter Semester 3/AIML/Project/Alzheiemer image/Alzheimer s Dataset/
 target_size=(224, 224),
    batch size=32,
    class_mode='binary', # binary classification
    subset='validation'
)
# Train the model with increased epochs
history=model.fit(train_generator, epochs=10,__
 ⇔validation_data=validation_generator)
Found 4098 images belonging to 4 classes.
Found 1023 images belonging to 4 classes.
Epoch 1/10
c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\site-
packages\keras\src\trainers\data_adapters\py_dataset_adapter.py:122:
UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in
its constructor. `**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will be
ignored.
 self._warn_if_super_not_called()
                   286s 2s/step -
accuracy: 0.4609 - loss: 1.2016 - val_accuracy: 0.4907 - val_loss: 1.0399
Epoch 2/10
129/129
                   285s 2s/step -
accuracy: 0.5348 - loss: 0.9941 - val_accuracy: 0.4213 - val_loss: 1.1297
Epoch 3/10
129/129
                   299s 2s/step -
accuracy: 0.5073 - loss: 0.9986 - val_accuracy: 0.4985 - val_loss: 1.1067
Epoch 4/10
129/129
                   286s 2s/step -
accuracy: 0.4944 - loss: 1.0155 - val_accuracy: 0.4936 - val_loss: 1.2130
Epoch 5/10
129/129
                   293s 2s/step -
accuracy: 0.5255 - loss: 0.9826 - val_accuracy: 0.3910 - val_loss: 1.1155
Epoch 6/10
129/129
                   296s 2s/step -
```

```
accuracy: 0.5347 - loss: 0.9494 - val_accuracy: 0.4936 - val_loss: 1.3215
     Epoch 7/10
     129/129
                         292s 2s/step -
     accuracy: 0.5414 - loss: 0.9495 - val_accuracy: 0.5249 - val_loss: 1.0896
     Epoch 8/10
     129/129
                         284s 2s/step -
     accuracy: 0.5404 - loss: 0.9491 - val_accuracy: 0.4976 - val_loss: 1.2133
     Epoch 9/10
     129/129
                         285s 2s/step -
     accuracy: 0.5413 - loss: 0.9352 - val_accuracy: 0.4946 - val_loss: 1.2045
     Epoch 10/10
     129/129
                         286s 2s/step -
     accuracy: 0.5481 - loss: 0.9118 - val_accuracy: 0.3822 - val_loss: 1.1716
[58]: from keras.models import Model
      def freeze_layers(model):
          for i in model.layers:
              i.trainable = False
              if isinstance(i, Model):
                  freeze_layers(i)
          return model
```

```
[59]: model_freezed = freeze_layers(model)
      model_freezed.save('alzheimer_res3.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')`.

[60]: model_freezed.summary()

Model: "functional_24"

Layer (type)	Output Shape	Param #	Connected to
<pre>input_layer_14 (InputLayer)</pre>	(None, 224, 224, 3)	0	-
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3)	0	input_layer_14[0
conv1_conv (Conv2D)	(None, 112, 112, 64)	9,472	conv1_pad[0][0]

conv1_bn (BatchNormalizatio	(None, 64)	112	, 112,	256	conv1_conv[0][0]
conv1_relu (Activation)	(None, 64)	112	, 112,	0	conv1_bn[0][0]
pool1_pad (ZeroPadding2D)	(None, 64)	114	, 114,	0	conv1_relu[0][0]
<pre>pool1_pool (MaxPooling2D)</pre>	(None, 64)	56,	56,	0	pool1_pad[0][0]
conv2_block1_1_conv (Conv2D)	(None, 64)	56,	56,	4,160	pool1_pool[0][0]
conv2_block1_1_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block1_1_c
<pre>conv2_block1_1_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block1_1_b
conv2_block1_2_conv (Conv2D)	(None, 64)	56,	56,	36,928	conv2_block1_1_r
conv2_block1_2_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block1_2_c
<pre>conv2_block1_2_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block1_2_b
conv2_block1_0_conv (Conv2D)	(None, 256)	56,	56,	16,640	pool1_pool[0][0]
conv2_block1_3_conv (Conv2D)	(None, 256)	56,	56,	16,640	conv2_block1_2_r
conv2_block1_0_bn (BatchNormalizatio	(None, 256)	56,	56,	1,024	conv2_block1_0_c
conv2_block1_3_bn (BatchNormalizatio	(None, 256)	56,	56,	1,024	conv2_block1_3_c
conv2_block1_add (Add)	(None, 256)	56,	56,	0	conv2_block1_0_b conv2_block1_3_b
conv2_block1_out (Activation)	(None, 256)	56,	56,	0	conv2_block1_add

conv2_block2_1_conv (Conv2D)	(None, 64)	56,	56,	16,448	conv2_block1_out
conv2_block2_1_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block2_1_c
<pre>conv2_block2_1_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block2_1_b
<pre>conv2_block2_2_conv (Conv2D)</pre>	(None, 64)	56,	56,	36,928	conv2_block2_1_r
conv2_block2_2_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block2_2_c
<pre>conv2_block2_2_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block2_2_b
<pre>conv2_block2_3_conv (Conv2D)</pre>	(None, 256)	56,	56,	16,640	conv2_block2_2_r
conv2_block2_3_bn (BatchNormalizatio	(None, 256)	56,	56,	1,024	conv2_block2_3_c
conv2_block2_add (Add)	(None, 256)	56,	56,	0	conv2_block1_out conv2_block2_3_b
<pre>conv2_block2_out (Activation)</pre>	(None, 256)	56,	56,	0	conv2_block2_add
<pre>conv2_block3_1_conv (Conv2D)</pre>	(None, 64)	56,	56,	16,448	conv2_block2_out
conv2_block3_1_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block3_1_c
<pre>conv2_block3_1_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block3_1_b
<pre>conv2_block3_2_conv (Conv2D)</pre>	(None, 64)	56,	56,	36,928	conv2_block3_1_r
conv2_block3_2_bn (BatchNormalizatio	(None, 64)	56,	56,	256	conv2_block3_2_c
<pre>conv2_block3_2_relu (Activation)</pre>	(None, 64)	56,	56,	0	conv2_block3_2_b

<pre>conv2_block3_3_conv (Conv2D)</pre>	(None, 256)	56,	56,	16,640	conv2_block3_2_r
conv2_block3_3_bn (BatchNormalizatio	(None, 256)	56,	56,	1,024	conv2_block3_3_c
conv2_block3_add (Add)	(None, 256)	56,	56,	0	conv2_block2_out conv2_block3_3_b
conv2_block3_out (Activation)	(None, 256)	56,	56,	0	conv2_block3_add
conv3_block1_1_conv (Conv2D)	(None, 128)	28,	28,	32,896	conv2_block3_out
conv3_block1_1_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block1_1_c
<pre>conv3_block1_1_relu (Activation)</pre>	(None, 128)	28,	28,	0	conv3_block1_1_b
conv3_block1_2_conv (Conv2D)	(None, 128)	28,	28,	147,584	conv3_block1_1_r
conv3_block1_2_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block1_2_c
conv3_block1_2_relu (Activation)	(None, 128)	28,	28,	0	conv3_block1_2_b
conv3_block1_0_conv (Conv2D)	(None, 512)	28,	28,	131,584	conv2_block3_out
conv3_block1_3_conv (Conv2D)	(None, 512)	28,	28,	66,048	conv3_block1_2_r
conv3_block1_0_bn (BatchNormalizatio	(None, 512)	28,	28,	2,048	conv3_block1_0_c
conv3_block1_3_bn (BatchNormalizatio	(None, 512)	28,	28,	2,048	conv3_block1_3_c
conv3_block1_add (Add)	(None, 512)	28,	28,	0	conv3_block1_0_b conv3_block1_3_b
conv3_block1_out (Activation)	(None, 512)	28,	28,	0	conv3_block1_add

conv3_block2_1_conv (Conv2D)	(None, 128)	28,	28,	65,664	conv3_block1_out
conv3_block2_1_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block2_1_c
<pre>conv3_block2_1_relu (Activation)</pre>	(None, 128)	28,	28,	0	conv3_block2_1_b
conv3_block2_2_conv (Conv2D)	(None, 128)	28,	28,	147,584	conv3_block2_1_r
conv3_block2_2_bn (BatchNormalizatio	(None,	28,	28,	512	conv3_block2_2_c
<pre>conv3_block2_2_relu (Activation)</pre>	(None, 128)	28,	28,	0	conv3_block2_2_b
conv3_block2_3_conv (Conv2D)	(None, 512)	28,	28,	66,048	conv3_block2_2_r
conv3_block2_3_bn (BatchNormalizatio	(None, 512)	28,	28,	2,048	conv3_block2_3_c
conv3_block2_add (Add)	(None, 512)	28,	28,	0	conv3_block1_out conv3_block2_3_b
conv3_block2_out (Activation)	(None, 512)	28,	28,	0	conv3_block2_add
conv3_block3_1_conv (Conv2D)	(None, 128)	28,	28,	65,664	conv3_block2_out
conv3_block3_1_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block3_1_c
conv3_block3_1_relu (Activation)	(None, 128)	28,	28,	0	conv3_block3_1_b
conv3_block3_2_conv (Conv2D)	(None, 128)	28,	28,	147,584	conv3_block3_1_r
conv3_block3_2_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block3_2_c
conv3_block3_2_relu (Activation)	(None, 128)	28,	28,	0	conv3_block3_2_b

<pre>conv3_block3_3_conv (Conv2D)</pre>	(None, 512)	28,	28,	66,048	conv3_block3_2_r
conv3_block3_3_bn (BatchNormalizatio	(None, 512)	28,	28,	2,048	conv3_block3_3_c
conv3_block3_add (Add)	(None, 512)	28,	28,	0	conv3_block2_out conv3_block3_3_b
<pre>conv3_block3_out (Activation)</pre>	(None, 512)	28,	28,	0	conv3_block3_add
<pre>conv3_block4_1_conv (Conv2D)</pre>	(None, 128)	28,	28,	65,664	conv3_block3_out
conv3_block4_1_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block4_1_c
<pre>conv3_block4_1_relu (Activation)</pre>	(None, 128)	28,	28,	0	conv3_block4_1_b
conv3_block4_2_conv (Conv2D)	(None, 128)	28,	28,	147,584	conv3_block4_1_r
conv3_block4_2_bn (BatchNormalizatio	(None, 128)	28,	28,	512	conv3_block4_2_c
<pre>conv3_block4_2_relu (Activation)</pre>	(None, 128)	28,	28,	0	conv3_block4_2_b
conv3_block4_3_conv (Conv2D)	(None, 512)	28,	28,	66,048	conv3_block4_2_r
conv3_block4_3_bn (BatchNormalizatio	(None, 512)	28,	28,	2,048	conv3_block4_3_c
conv3_block4_add (Add)	(None, 512)	28,	28,	0	conv3_block3_out conv3_block4_3_b
conv3_block4_out (Activation)	(None, 512)	28,	28,	0	conv3_block4_add
<pre>conv4_block1_1_conv (Conv2D)</pre>	(None, 256)	14,	14,	131,328	conv3_block4_out
conv4_block1_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block1_1_c

<pre>conv4_block1_1_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block1_1_b
conv4_block1_2_conv (Conv2D)	(None, 256)	14,	14,	590,080	conv4_block1_1_r
conv4_block1_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block1_2_c
<pre>conv4_block1_2_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block1_2_b
conv4_block1_0_conv (Conv2D)	(None, 1024)	14,	14,	525,312	conv3_block4_out
conv4_block1_3_conv (Conv2D)	(None, 1024)	14,	14,	263,168	conv4_block1_2_r
conv4_block1_0_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block1_0_c
conv4_block1_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block1_3_c
conv4_block1_add (Add)	(None, 1024)	14,	14,	0	conv4_block1_0_b conv4_block1_3_b
conv4_block1_out (Activation)	(None, 1024)	14,	14,	0	conv4_block1_add
conv4_block2_1_conv (Conv2D)	(None, 256)	14,	14,	262,400	conv4_block1_out
conv4_block2_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block2_1_c
conv4_block2_1_relu (Activation)	(None, 256)	14,	14,	0	conv4_block2_1_b
conv4_block2_2_conv (Conv2D)	(None, 256)	14,	14,	590,080	conv4_block2_1_r
conv4_block2_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block2_2_c
conv4_block2_2_relu (Activation)	(None, 256)	14,	14,	0	conv4_block2_2_b

<pre>conv4_block2_3_conv (Conv2D)</pre>	(None, 1024)	14,	14,	263,168	conv4_block2_2_r
conv4_block2_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block2_3_c
conv4_block2_add (Add)	(None, 1024)	14,	14,	0	conv4_block1_out conv4_block2_3_b
conv4_block2_out (Activation)	(None, 1024)	14,	14,	0	conv4_block2_add
<pre>conv4_block3_1_conv (Conv2D)</pre>	(None, 256)	14,	14,	262,400	conv4_block2_out
conv4_block3_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block3_1_c
<pre>conv4_block3_1_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block3_1_b
conv4_block3_2_conv (Conv2D)	(None, 256)	14,	14,	590,080	conv4_block3_1_r
conv4_block3_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block3_2_c
<pre>conv4_block3_2_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block3_2_b
conv4_block3_3_conv (Conv2D)	(None, 1024)	14,	14,	263,168	conv4_block3_2_r
conv4_block3_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block3_3_c
conv4_block3_add (Add)	(None, 1024)	14,	14,	0	conv4_block2_out conv4_block3_3_b
<pre>conv4_block3_out (Activation)</pre>	(None, 1024)	14,	14,	0	conv4_block3_add
conv4_block4_1_conv (Conv2D)	(None, 256)	14,	14,	262,400	conv4_block3_out
conv4_block4_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block4_1_c

<pre>conv4_block4_1_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block4_1_b
conv4_block4_2_conv (Conv2D)	(None, 256)	14,	14,	590,080	conv4_block4_1_r
conv4_block4_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block4_2_c
<pre>conv4_block4_2_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block4_2_b
conv4_block4_3_conv (Conv2D)	(None, 1024)	14,	14,	263,168	conv4_block4_2_r
conv4_block4_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block4_3_c
conv4_block4_add (Add)	(None, 1024)	14,	14,	0	conv4_block3_out conv4_block4_3_b
conv4_block4_out (Activation)	(None, 1024)	14,	14,	0	conv4_block4_add
conv4_block5_1_conv (Conv2D)	(None, 256)	14,	14,	262,400	conv4_block4_out
conv4_block5_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block5_1_c
<pre>conv4_block5_1_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block5_1_b
conv4_block5_2_conv (Conv2D)	(None, 256)	14,	14,	590,080	conv4_block5_1_r
conv4_block5_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block5_2_c
<pre>conv4_block5_2_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block5_2_b
conv4_block5_3_conv (Conv2D)	(None, 1024)	14,	14,	263,168	conv4_block5_2_r
conv4_block5_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block5_3_c

conv4_block5_add (Add)	(None, 1024)	14,	14,	0	conv4_block4_out conv4_block5_3_b
<pre>conv4_block5_out (Activation)</pre>	(None, 1024)	14,	14,	0	conv4_block5_add
<pre>conv4_block6_1_conv (Conv2D)</pre>	(None, 256)	14,	14,	262,400	conv4_block5_out
conv4_block6_1_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block6_1_c
<pre>conv4_block6_1_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block6_1_b
<pre>conv4_block6_2_conv (Conv2D)</pre>	(None, 256)	14,	14,	590,080	conv4_block6_1_r
conv4_block6_2_bn (BatchNormalizatio	(None, 256)	14,	14,	1,024	conv4_block6_2_c
<pre>conv4_block6_2_relu (Activation)</pre>	(None, 256)	14,	14,	0	conv4_block6_2_b
<pre>conv4_block6_3_conv (Conv2D)</pre>	(None, 1024)	14,	14,	263,168	conv4_block6_2_r
conv4_block6_3_bn (BatchNormalizatio	(None, 1024)	14,	14,	4,096	conv4_block6_3_c
conv4_block6_add (Add)	(None, 1024)	14,	14,	0	conv4_block5_out conv4_block6_3_b
conv4_block6_out (Activation)	(None, 1024)	14,	14,	0	conv4_block6_add
<pre>conv5_block1_1_conv (Conv2D)</pre>	(None,	7, 7	7, 512)	524,800	conv4_block6_out
conv5_block1_1_bn (BatchNormalizatio	(None,	7, 7	7, 512)	2,048	conv5_block1_1_c
<pre>conv5_block1_1_relu (Activation)</pre>	(None,	7, 7	7, 512)	0	conv5_block1_1_b
conv5_block1_2_conv (Conv2D)	(None,	7, 7	7, 512)	2,359,808	conv5_block1_1_r

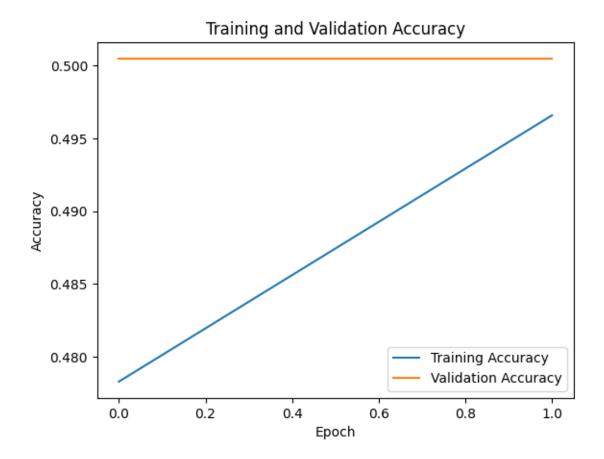
conv5_block1_2_bn (BatchNormalizatio	(None,	7,	7,	512)	2,048	conv5_block1_2_c
<pre>conv5_block1_2_relu (Activation)</pre>	(None,	7,	7,	512)	0	conv5_block1_2_b
<pre>conv5_block1_0_conv (Conv2D)</pre>	(None, 2048)	7,	7,		2,099,200	conv4_block6_out
<pre>conv5_block1_3_conv (Conv2D)</pre>	(None, 2048)	7,	7,		1,050,624	conv5_block1_2_r
conv5_block1_0_bn (BatchNormalizatio	(None, 2048)	7,	7,		8,192	conv5_block1_0_c
conv5_block1_3_bn (BatchNormalizatio	(None, 2048)	7,	7,		8,192	conv5_block1_3_c
conv5_block1_add (Add)	(None, 2048)	7,	7,		0	conv5_block1_0_b conv5_block1_3_b
<pre>conv5_block1_out (Activation)</pre>	(None, 2048)	7,	7,		0	conv5_block1_add
<pre>conv5_block2_1_conv (Conv2D)</pre>	(None,	7,	7,	512)	1,049,088	conv5_block1_out
conv5_block2_1_bn (BatchNormalizatio	(None,	7,	7,	512)	2,048	conv5_block2_1_c
<pre>conv5_block2_1_relu (Activation)</pre>	(None,	7,	7,	512)	0	conv5_block2_1_b
<pre>conv5_block2_2_conv (Conv2D)</pre>	(None,	7,	7,	512)	2,359,808	conv5_block2_1_r
conv5_block2_2_bn (BatchNormalizatio	(None,	7,	7,	512)	2,048	conv5_block2_2_c
<pre>conv5_block2_2_relu (Activation)</pre>	(None,	7,	7,	512)	0	conv5_block2_2_b
conv5_block2_3_conv (Conv2D)	(None, 2048)	7,	7,		1,050,624	conv5_block2_2_r
conv5_block2_3_bn (BatchNormalizatio	(None, 2048)	7,	7,		8,192	conv5_block2_3_c

conv5_block2_add (Add)	(None, 7, 7, 2048)	0	conv5_block1_out conv5_block2_3_b
conv5_block2_out (Activation)	(None, 7, 7, 2048)	0	conv5_block2_add
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1,049,088	conv5_block2_out
conv5_block3_1_bn (BatchNormalizatio	(None, 7, 7, 512)	2,048	conv5_block3_1_c
<pre>conv5_block3_1_relu (Activation)</pre>	(None, 7, 7, 512)	0	conv5_block3_1_b
conv5_block3_2_conv (Conv2D)	(None, 7, 7, 512)	2,359,808	conv5_block3_1_r
conv5_block3_2_bn (BatchNormalizatio	(None, 7, 7, 512)	2,048	conv5_block3_2_c
conv5_block3_2_relu (Activation)	(None, 7, 7, 512)	0	conv5_block3_2_b
conv5_block3_3_conv (Conv2D)	(None, 7, 7, 2048)	1,050,624	conv5_block3_2_r
conv5_block3_3_bn (BatchNormalizatio	(None, 7, 7, 2048)	8,192	conv5_block3_3_c
conv5_block3_add (Add)	(None, 7, 7, 2048)	0	conv5_block2_out conv5_block3_3_b
conv5_block3_out (Activation)	(None, 7, 7, 2048)	0	conv5_block3_add
flatten_14 (Flatten)	(None, 100352)	0	conv5_block3_out
dense_38 (Dense)	(None, 256)	25,690,368	flatten_14[0][0]
dense_39 (Dense)	(None, 128)	32,896	dense_38[0][0]
dense_40 (Dense)	(None, 4)	516	dense_39[0][0]

```
Total params: 109,690,382 (418.44 MB)
      Trainable params: 0 (0.00 B)
      Non-trainable params: 49,311,492 (188.11 MB)
      Optimizer params: 60,378,890 (230.33 MB)
[61]: test_model=load_model(r"C:\Users\harsa\Downloads\alzheimer_res3.h5")
     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.
[62]: img=image.load_img(r"D:\Winter Semester_
       -3\AIML\Project\Alzheiemer_image\Alzheimer s Dataset\test\NonDemented\26 (62).
       →jpg")
      img=img.resize((180,180))
[63]: x=image.img_to_array(img)
      x=np.expand_dims(x,axis=0)
      img_data=preprocess_input(x)
[64]: from tensorflow.image import resize
      # Resize input data to match the model's input shape (224x224)
      x_{resized} = resize(x, (224, 224))
      # Preprocess input data if needed (e.g., normalize, scale)
      x_preprocessed = preprocess_input(x_resized)
      # Make predictions using the preprocessed input data
      predictions = test_model.predict(x_preprocessed)
      print(np.argmax(predictions, axis=1)[0]) # Example: Get the predicted class
                     2s 2s/step
     1/1
     2
[67]: import numpy as np
      from sklearn.metrics import classification_report
      import matplotlib.pyplot as plt
      test generator = train datagen.flow from directory(
          "D:/Winter Semester 3/AIML/Project/Alzheiemer_image/Alzheimer_s Dataset/
       target_size=(224, 224),
```

```
batch_size=32,
    class_mode='binary', # binary classification
    subset='training'
validation_generator = train_datagen.flow_from_directory(
    "D:/Winter Semester 3/AIML/Project/Alzheiemer_image/Alzheimer_s Dataset/
 target_size=(224, 224),
    batch_size=32,
    class_mode='binary', # binary classification
    subset='validation'
)
# Train the model
#history = model.fit(train_generator, epochs=2,__
 ⇒validation_data=validation_generator)
# Plot training and validation accuracy
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val accuracy'], label='Validation Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.title('Training and Validation Accuracy')
plt.show()
# Evaluate the model on test data
loss, accuracy = model.evaluate(test generator)
print(f"Test Loss: {loss}")
print(f"Test Accuracy: {accuracy}")
# Generate predictions for test data
predictions = model.predict(test generator)
predicted_classes = np.argmax(predictions, axis=1)
# Get true classes from the generator
true_classes = test_generator.classes
class_labels = list(test_generator.class_indices.keys())
# Generate classification report
report = classification_report(true_classes, predicted_classes,_
 →target_names=class_labels)
print("Classification Report:\n", report)
```

Found 1025 images belonging to 4 classes. Found 254 images belonging to 4 classes.



c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\sitepackages\keras\src\trainers\data_adapters\py_dataset_adapter.py:122:
UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in
its constructor. `**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will be
ignored.

self._warn_if_super_not_called()

33/33 47s 1s/step accuracy: 0.4869 - loss: 1.1885
Test Loss: 1.1624151468276978
Test Accuracy: 0.4995121955871582
33/33 51s 1s/step

Classification Report:

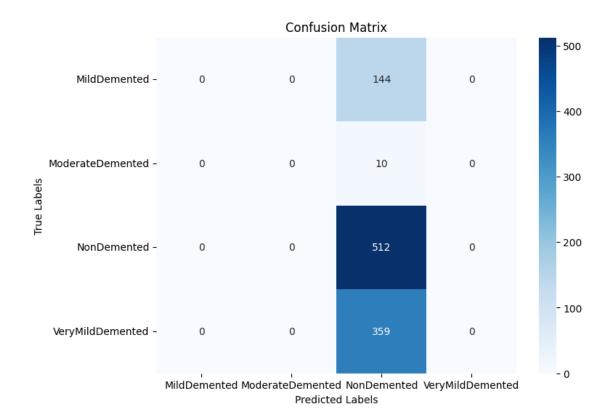
	precision	recall	f1-score	support
MildDemented	0.00	0.00	0.00	144
ModerateDemented	0.00	0.00	0.00	10
NonDemented	0.50	1.00	0.67	512
${\tt VeryMildDemented}$	0.00	0.00	0.00	359

```
      accuracy
      0.50
      1025

      macro avg
      0.12
      0.25
      0.17
      1025

      weighted avg
      0.25
      0.50
      0.33
      1025
```

```
c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\site-
     packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning:
     Precision and F-score are ill-defined and being set to 0.0 in labels with no
     predicted samples. Use `zero_division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
     c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\site-
     packages\sklearn\metrics\ classification.py:1471: UndefinedMetricWarning:
     Precision and F-score are ill-defined and being set to 0.0 in labels with no
     predicted samples. Use `zero_division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
     c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\site-
     packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning:
     Precision and F-score are ill-defined and being set to 0.0 in labels with no
     predicted samples. Use `zero_division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
[68]: import numpy as np
      from sklearn.metrics import classification_report, confusion_matrix
      import matplotlib.pyplot as plt
      import seaborn as sns
[69]: # Generate confusion matrix
      conf matrix = confusion matrix(true classes, predicted classes)
      # Plot confusion matrix
      plt.figure(figsize=(8, 6))
      sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues',__
       →xticklabels=class_labels, yticklabels=class_labels)
      plt.xlabel('Predicted Labels')
      plt.ylabel('True Labels')
      plt.title('Confusion Matrix')
      plt.show()
```



```
[70]: # Load test data
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(
    "D:/Winter Semester 3/AIML/Project/Alzheiemer_image/Alzheimer_s Dataset/
    test",
    target_size=(224, 224),
    batch_size=32,
    class_mode='binary' # binary classification
)

# Evaluate the model on test data
loss, accuracy = model.evaluate(test_generator)

print(f"Test Loss: {loss}")
print(f"Test Accuracy: {accuracy}")
```

Found 1279 images belonging to 4 classes.

```
c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\site-
packages\keras\src\trainers\data_adapters\py_dataset_adapter.py:122:
UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in
its constructor. `**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will be
```

40/40 55s 1s/step Classification Report:

	precision	recall	f1-score	support
MildDemented	0.00	0.00	0.00	179
${\tt ModerateDemented}$	0.00	0.00	0.00	12
NonDemented	0.50	1.00	0.67	640
VeryMildDemented	0.00	0.00	0.00	448
accuracy			0.50	1279
macro avg	0.13	0.25	0.17	1279
weighted avg	0.25	0.50	0.33	1279

c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\sitepackages\sklearn\metrics_classification.py:1471: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\sitepackages\sklearn\metrics_classification.py:1471: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

c:\Users\harsa\AppData\Local\Programs\Python\Python310\lib\sitepackages\sklearn\metrics_classification.py:1471: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. _warn_prf(average, modifier, msg_start, len(result))