## Personal Project\_04\_v10\_test1\_4conv-layer\_run78\_very advanced control 4\_autorun

May 7, 2025

```
[1]: from tensorflow.keras.callbacks import LearningRateScheduler
     from sklearn.metrics import classification_report, confusion_matrix
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     %matplotlib inline
     import matplotlib.image as mpimg
     import tensorflow as tf
     import os
     class EarlyStoppingCallback(tf.keras.callbacks.Callback):
         def on_epoch_end(self, epoch, logs=None):
             train_accuracy = logs.get('accuracy')
             val_accuracy = logs.get('val_accuracy')
             if train_accuracy >= desired_train_accuracy and val_accuracy >=_
      →desired_val_accuracy:
                 self.model.stop_training = True
                 print("Reached desired accuracy so cancelling training!")
     # target accuracy values:
     desired_train_accuracy = 0.89
     desired_val_accuracy = 0.89
     # maximum trial number:
     trial num = 50
     # maximum possible epoch:
     epochs = 20
     TRAIN_ACC=0.1
     VAL_ACC=0.1
     try_num = 1
     condition = True
```

```
while (try_num<trial_num and condition==True):</pre>
    # DOE factors:
    learning_rate = 0.0005
    dropout_value = 0.2
    \# n\text{-}conv\_layers = 4
    n_units_last_layer = 4096
    n filters 11 = 32
    n_filters_12 = 64
    # other factors:
    img_size = 130
    batch_size = 32
    validation_split = 0.1 # 10% for validation
    test_split = 0.00 # 0% for testing
    shuffle_buffer_size = 1000
    seed_num = 101
    desired\_accuracy = 0.99 # it should be active if EarlyStoppingCallback is
 \rightarrowactivated
    loss = 'binary crossentropy'
    #optimizer = tf.keras.optimizers.RMSprop(learning_rate=learning_rate)
    optimizer = tf.keras.optimizers.Adam(learning_rate=learning_rate)
    metrics = ['accuracy']
    f_mode = 'nearest' # fill_mode in image augmentation
    \#DATA\_DIR = "D: \CS on line courses \Free DataSets \Free Images \Easier
 ⇔portrait images_GPU_03"
    DATA_DIR = "/Users/hossein/Downloads/Easier portrait images_GPU_03"
    # Subdirectories for each class
    data dir woman = os.path.join(DATA DIR, 'woman')
    data_dir_man = os.path.join(DATA_DIR, 'man')
    image_size = (img_size, img_size) # Resize images to this size
    # Load train dataset (excluding validation & test set):
    train_dataset = tf.keras.utils.image_dataset_from_directory(
        directory = DATA_DIR,
        image_size = image_size,
        batch_size = batch_size,
        label_mode='binary',
        validation_split = validation_split + test_split, # Total split for_
 ⇔val + test
        subset = "training",
```

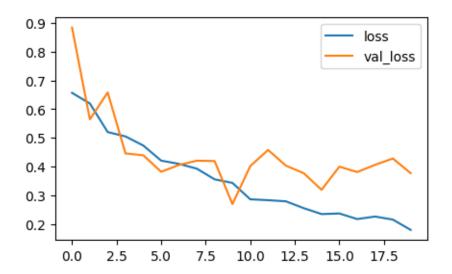
```
seed = seed_num
  )
  # Load validation dataset
  val_dataset = tf.keras.utils.image_dataset_from_directory(
      directory = DATA_DIR,
      image_size = image_size,
      batch_size = batch_size,
      label mode='binary',
      validation_split = validation_split + test_split,
      subset = "validation",
      seed = seed_num
  )
  # Further manually split validation dataset to extract test dataset
  val_batches = tf.data.experimental.cardinality(val_dataset)
  # Compute test dataset size (number of batches)
  test_size = round(val_batches.numpy() * (test_split / (validation_split + L
→test_split)))
  # Split validation dataset into validation and test subsets
  test dataset = val dataset.take(test size)
  val_dataset = val_dataset.skip(test_size)
  # Optimize for performance
  AUTOTUNE = tf.data.AUTOTUNE
  training_dataset = train_dataset.cache().shuffle(shuffle_buffer_size).
→prefetch(buffer_size = AUTOTUNE)
  validation_dataset = val_dataset.cache().prefetch(buffer_size = AUTOTUNE)
  test_dataset = test_dataset.cache().prefetch(buffer_size = AUTOTUNE)
  # Get the first batch of images and labels
  for images, labels in training_dataset.take(1):
          example_batch_images = images
          example_batch_labels = labels
  max_pixel = np.max(example_batch_images)
  def scheduler(epoch, lr):
      if epoch < 10:</pre>
           if epoch % 5 == 0 and epoch > 0:
              return lr / 1
          return lr
      elif epoch < 15:
          if epoch % 5 == 0 and epoch > 0:
              return lr / 2
          return lr
      elif epoch < 30:</pre>
```

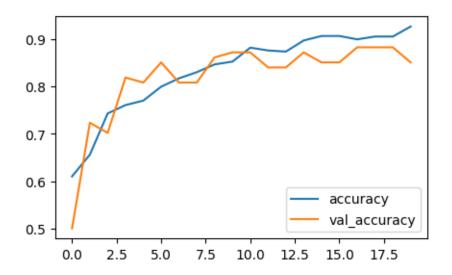
```
if epoch \% 5 == 0 and epoch > 0:
               return lr / 1
           return lr
       return lr
  lr_callback = LearningRateScheduler(scheduler)
  # augmentation_model
  def augment_model():
       augmentation_model = tf.keras.Sequential([
           # Specify the input shape.
           tf.keras.Input(shape = (img_size, img_size, 3)),
           tf.keras.layers.RandomFlip("horizontal"),
           tf.keras.layers.RandomRotation(0.1, fill_mode = f_mode),
           #tf.keras.layers.RandomTranslation(0.1, 0.1, fill_mode = f_mode),
           #tf.keras.layers.RandomZoom(0.1, fill_mode=f_mode)
           ])
      return augmentation_model
  def create_and_compile_model():
       augmentation_layers = augment_model()
      model = tf.keras.Sequential([
           # Note: the input shape is the desired size of the image: 150x150_{\square}
⇔with 3 bytes for color
           tf.keras.layers.InputLayer(shape = (img_size, img_size, 3)),
           augmentation_layers,
           tf.keras.layers.Rescaling(1./255),
                    CONV_LAYER_1:
                                      #####
           tf.keras.layers.Conv2D(n_filters_11, (4, 4), activation = 'linear'),
           tf.keras.layers.MaxPooling2D(2, 2),
                    CONV LAYER 2:
           #####
           tf.keras.layers.Conv2D(n_filters_12, (3, 3), activation = 'relu'),
           tf.keras.layers.MaxPooling2D(2, 2),
                    CONV_LAYER_3:
           #####
                                      #####
           tf.keras.layers.Conv2D(64, (3, 3), activation = 'relu'),
           tf.keras.layers.MaxPooling2D(2, 2),
           #####
                    CONV_LAYER_4:
           tf.keras.layers.Conv2D(64, (3, 3), activation = 'relu'),
           tf.keras.layers.MaxPooling2D(2, 2),
           tf.keras.layers.Flatten(),
           tf.keras.layers.Dropout(dropout_value),
                    BEFORE_LAST_LAYER:
                                           #####
           tf.keras.layers.Dense(n_units_last_layer, activation = 'relu'),
           # It will contain a value from O-1 where O for the class 'female' \Box
⇔and 1 for the 'male'
           tf.keras.layers.Dense(1, activation = 'sigmoid')])
      model.compile(
           loss = loss,
```

```
optimizer = optimizer,
          metrics = metrics
      return model
  # Create the compiled but untrained model
  def reset_weights(model):
      for layer in model.layers:
          if hasattr(layer, 'kernel_initializer'):
              layer.kernel.assign(layer.kernel initializer(layer.kernel.
⇔shape))
          if hasattr(layer, 'bias_initializer'):
              layer.bias.assign(layer.bias_initializer(layer.bias.shape))
  model = create_and_compile_model()
  reset_weights(model) # Reset all layer weights
  training_history = model.fit(training_dataset,
                               epochs=epochs,
                               validation_data=validation_dataset,
                                callbacks=[lr callback,
→EarlyStoppingCallback()],
                               verbose=1)
  result_history = pd.DataFrame(model.history.history)
  TRAIN_ACC = result_history['accuracy'].iloc[-1]
  print(f"Current training accuracy: {TRAIN_ACC}")
  VAL ACC = result history['val accuracy'].iloc[-1]
  print(f"Current validation accuracy: {VAL ACC}")
  # Restart script
  print("Reseting all weights...")
  print(f'Current number of trials: {try_num}')
  try num += 1
  result_history[['loss', 'val_loss']].plot(figsize=(5, 3))
  result_history[['accuracy', 'val_accuracy']].plot(figsize=(5, 3))
  plt.show()
  print(model.metrics_names)
  print(model.evaluate(validation_dataset))
  y_true = np.concatenate([y.numpy() for _, y in validation dataset])
  y_pred_prob = model.predict(validation_dataset)
  # Convert probabilities to class labels (0:Female or 1:Male)
  y_pred = (y_pred_prob > 0.5).astype(int).flatten()
  print("Classification Report:\n", classification_report(y_true, y_pred,_
→target_names=['Female', 'Male']))
  if (TRAIN_ACC>=desired_train_accuracy and VAL_ACC>=desired_val_accuracy):
      condition = False
      model.save('trained_model_run78_very_advanced_control.h5')
```

```
result_history.head(15)
Found 943 files belonging to 2 classes.
Using 849 files for training.
Found 943 files belonging to 2 classes.
Using 94 files for validation.
Epoch 1/20
2025-05-07 12:22:49.495620: I tensorflow/core/framework/local_rendezvous.cc:405]
Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
27/27
                  3s 98ms/step -
accuracy: 0.5822 - loss: 0.6885 - val_accuracy: 0.5000 - val_loss: 0.8843 -
learning_rate: 5.0000e-04
Epoch 2/20
27/27
                  3s 95ms/step -
accuracy: 0.6371 - loss: 0.6461 - val_accuracy: 0.7234 - val_loss: 0.5644 -
learning_rate: 5.0000e-04
Epoch 3/20
27/27
                  3s 98ms/step -
accuracy: 0.7372 - loss: 0.5271 - val_accuracy: 0.7021 - val_loss: 0.6582 -
learning_rate: 5.0000e-04
Epoch 4/20
27/27
                  3s 95ms/step -
accuracy: 0.7479 - loss: 0.5298 - val_accuracy: 0.8191 - val_loss: 0.4458 -
learning_rate: 5.0000e-04
Epoch 5/20
27/27
                  3s 97ms/step -
accuracy: 0.7642 - loss: 0.4580 - val_accuracy: 0.8085 - val_loss: 0.4392 -
learning_rate: 5.0000e-04
Epoch 6/20
27/27
                  3s 96ms/step -
accuracy: 0.7932 - loss: 0.4415 - val_accuracy: 0.8511 - val_loss: 0.3819 -
learning_rate: 5.0000e-04
Epoch 7/20
27/27
                  3s 97ms/step -
accuracy: 0.8219 - loss: 0.4030 - val_accuracy: 0.8085 - val_loss: 0.4056 -
learning_rate: 5.0000e-04
Epoch 8/20
27/27
                  3s 96ms/step -
accuracy: 0.8265 - loss: 0.4040 - val_accuracy: 0.8085 - val_loss: 0.4204 -
learning_rate: 5.0000e-04
Epoch 9/20
27/27
                  3s 96ms/step -
accuracy: 0.8598 - loss: 0.3465 - val_accuracy: 0.8617 - val_loss: 0.4194 -
learning_rate: 5.0000e-04
Epoch 10/20
```

```
27/27
                  3s 96ms/step -
accuracy: 0.8604 - loss: 0.3343 - val_accuracy: 0.8723 - val_loss: 0.2692 -
learning_rate: 5.0000e-04
Epoch 11/20
27/27
                  3s 97ms/step -
accuracy: 0.8968 - loss: 0.2748 - val_accuracy: 0.8723 - val_loss: 0.4017 -
learning rate: 2.5000e-04
Epoch 12/20
27/27
                  3s 96ms/step -
accuracy: 0.8852 - loss: 0.2814 - val_accuracy: 0.8404 - val_loss: 0.4579 -
learning_rate: 2.5000e-04
Epoch 13/20
27/27
                  3s 96ms/step -
accuracy: 0.8547 - loss: 0.3136 - val_accuracy: 0.8404 - val_loss: 0.4032 -
learning_rate: 2.5000e-04
Epoch 14/20
27/27
                  3s 96ms/step -
accuracy: 0.9058 - loss: 0.2361 - val_accuracy: 0.8723 - val_loss: 0.3769 -
learning_rate: 2.5000e-04
Epoch 15/20
27/27
                  3s 97ms/step -
accuracy: 0.8999 - loss: 0.2315 - val_accuracy: 0.8511 - val_loss: 0.3192 -
learning_rate: 2.5000e-04
Epoch 16/20
27/27
                  3s 97ms/step -
accuracy: 0.9107 - loss: 0.2446 - val_accuracy: 0.8511 - val_loss: 0.3997 -
learning_rate: 2.5000e-04
Epoch 17/20
27/27
                  3s 98ms/step -
accuracy: 0.9136 - loss: 0.2122 - val_accuracy: 0.8830 - val_loss: 0.3810 -
learning_rate: 2.5000e-04
Epoch 18/20
27/27
                  3s 97ms/step -
accuracy: 0.9086 - loss: 0.2189 - val_accuracy: 0.8830 - val_loss: 0.4060 -
learning rate: 2.5000e-04
Epoch 19/20
27/27
                  3s 97ms/step -
accuracy: 0.9036 - loss: 0.2123 - val_accuracy: 0.8830 - val_loss: 0.4281 -
learning_rate: 2.5000e-04
Epoch 20/20
27/27
                 3s 97ms/step -
accuracy: 0.9264 - loss: 0.1941 - val_accuracy: 0.8511 - val_loss: 0.3769 -
learning_rate: 2.5000e-04
Current training accuracy: 0.9269729256629944
Current validation accuracy: 0.8510638475418091
Reseting all weights...
Current number of trials: 1
```





[0.37694793939590454, 0.8510638475418091]

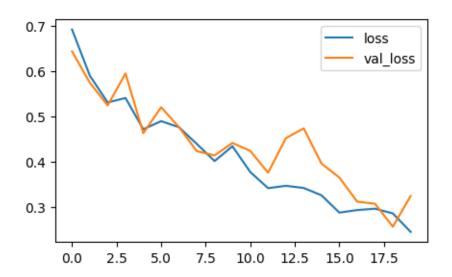
3/3 0s 36ms/step

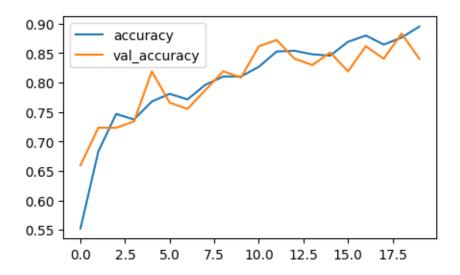
Classification Report:

	precision	recall	f1-score	support
Female Male	0.81 0.88	0.85 0.85	0.83 0.87	41 53
accuracy			0.85	94

```
0.85
                             0.85
                                       0.85
                                                   94
  macro avg
                   0.85
                                       0.85
                                                   94
weighted avg
                             0.85
Found 943 files belonging to 2 classes.
Using 849 files for training.
Found 943 files belonging to 2 classes.
Using 94 files for validation.
2025-05-07 12:23:42.466022: I tensorflow/core/framework/local_rendezvous.cc:405]
Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
Epoch 1/20
27/27
                  3s 100ms/step -
accuracy: 0.5201 - loss: 0.7306 - val_accuracy: 0.6596 - val_loss: 0.6441 -
learning_rate: 5.0000e-04
Epoch 2/20
27/27
                  3s 96ms/step -
accuracy: 0.6903 - loss: 0.5974 - val_accuracy: 0.7234 - val_loss: 0.5745 -
learning_rate: 5.0000e-04
Epoch 3/20
27/27
                  3s 98ms/step -
accuracy: 0.7712 - loss: 0.5243 - val_accuracy: 0.7234 - val_loss: 0.5244 -
learning rate: 5.0000e-04
Epoch 4/20
27/27
                  3s 96ms/step -
accuracy: 0.7126 - loss: 0.5493 - val_accuracy: 0.7340 - val_loss: 0.5952 -
learning_rate: 5.0000e-04
Epoch 5/20
27/27
                  3s 96ms/step -
accuracy: 0.7730 - loss: 0.4587 - val_accuracy: 0.8191 - val_loss: 0.4628 -
learning_rate: 5.0000e-04
Epoch 6/20
27/27
                  3s 96ms/step -
accuracy: 0.7853 - loss: 0.4603 - val_accuracy: 0.7660 - val_loss: 0.5203 -
learning_rate: 5.0000e-04
Epoch 7/20
27/27
                  3s 96ms/step -
accuracy: 0.7849 - loss: 0.4686 - val accuracy: 0.7553 - val loss: 0.4762 -
learning_rate: 5.0000e-04
Epoch 8/20
27/27
                  3s 96ms/step -
accuracy: 0.7938 - loss: 0.4376 - val_accuracy: 0.7872 - val_loss: 0.4229 -
learning_rate: 5.0000e-04
Epoch 9/20
                  3s 100ms/step -
27/27
accuracy: 0.8030 - loss: 0.4236 - val_accuracy: 0.8191 - val_loss: 0.4136 -
learning_rate: 5.0000e-04
Epoch 10/20
27/27
                  3s 96ms/step -
```

```
accuracy: 0.8022 - loss: 0.4394 - val_accuracy: 0.8085 - val_loss: 0.4412 -
learning_rate: 5.0000e-04
Epoch 11/20
27/27
                  3s 97ms/step -
accuracy: 0.8375 - loss: 0.3721 - val_accuracy: 0.8617 - val_loss: 0.4235 -
learning_rate: 2.5000e-04
Epoch 12/20
27/27
                  3s 96ms/step -
accuracy: 0.8414 - loss: 0.3441 - val_accuracy: 0.8723 - val_loss: 0.3752 -
learning_rate: 2.5000e-04
Epoch 13/20
27/27
                  3s 96ms/step -
accuracy: 0.8530 - loss: 0.3435 - val_accuracy: 0.8404 - val_loss: 0.4519 -
learning_rate: 2.5000e-04
Epoch 14/20
27/27
                  3s 96ms/step -
accuracy: 0.8327 - loss: 0.3620 - val_accuracy: 0.8298 - val_loss: 0.4735 -
learning_rate: 2.5000e-04
Epoch 15/20
27/27
                  3s 98ms/step -
accuracy: 0.8462 - loss: 0.3397 - val_accuracy: 0.8511 - val_loss: 0.3953 -
learning_rate: 2.5000e-04
Epoch 16/20
27/27
                  3s 97ms/step -
accuracy: 0.8662 - loss: 0.2780 - val_accuracy: 0.8191 - val_loss: 0.3644 -
learning_rate: 2.5000e-04
Epoch 17/20
27/27
                  3s 97ms/step -
accuracy: 0.8829 - loss: 0.3070 - val_accuracy: 0.8617 - val_loss: 0.3110 -
learning_rate: 2.5000e-04
Epoch 18/20
27/27
                  3s 97ms/step -
accuracy: 0.8687 - loss: 0.2847 - val accuracy: 0.8404 - val loss: 0.3064 -
learning_rate: 2.5000e-04
Epoch 19/20
27/27
                  3s 97ms/step -
accuracy: 0.8816 - loss: 0.2991 - val accuracy: 0.8830 - val loss: 0.2557 -
learning_rate: 2.5000e-04
Epoch 20/20
27/27
                  3s 97ms/step -
accuracy: 0.8928 - loss: 0.2598 - val_accuracy: 0.8404 - val_loss: 0.3236 -
learning_rate: 2.5000e-04
Current training accuracy: 0.8951708078384399
Current validation accuracy: 0.8404255509376526
Reseting all weights...
Current number of trials: 2
```





[0.32360365986824036, 0.8404255509376526]

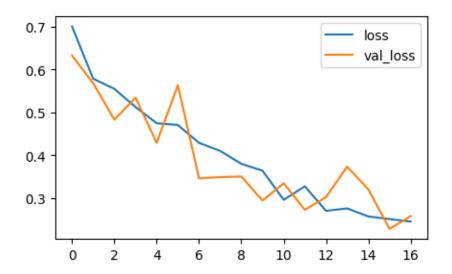
3/3 0s 37ms/step

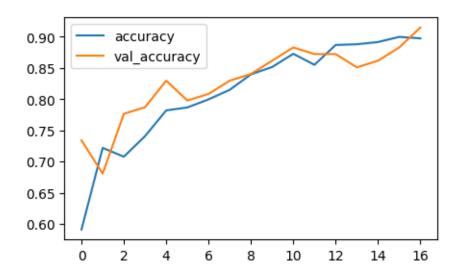
Classification Report:

	precision	recall	f1-score	support
Female	0.78	0.88	0.83	41
Male	0.90	0.81	0.85	53
accuracy			0.84	94

```
0.84
                             0.84
                                       0.84
                                                   94
  macro avg
                   0.85
                                       0.84
                                                   94
weighted avg
                             0.84
Found 943 files belonging to 2 classes.
Using 849 files for training.
Found 943 files belonging to 2 classes.
Using 94 files for validation.
2025-05-07 12:24:35.960210: I tensorflow/core/framework/local_rendezvous.cc:405]
Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
Epoch 1/20
27/27
                  3s 102ms/step -
accuracy: 0.5591 - loss: 0.7182 - val_accuracy: 0.7340 - val_loss: 0.6327 -
learning_rate: 5.0000e-04
Epoch 2/20
27/27
                  3s 97ms/step -
accuracy: 0.7219 - loss: 0.5909 - val_accuracy: 0.6809 - val_loss: 0.5680 -
learning_rate: 5.0000e-04
Epoch 3/20
27/27
                  3s 97ms/step -
accuracy: 0.6785 - loss: 0.5778 - val_accuracy: 0.7766 - val_loss: 0.4832 -
learning rate: 5.0000e-04
Epoch 4/20
27/27
                  3s 97ms/step -
accuracy: 0.7385 - loss: 0.5046 - val_accuracy: 0.7872 - val_loss: 0.5345 -
learning_rate: 5.0000e-04
Epoch 5/20
27/27
                  3s 97ms/step -
accuracy: 0.7804 - loss: 0.4804 - val_accuracy: 0.8298 - val_loss: 0.4290 -
learning_rate: 5.0000e-04
Epoch 6/20
27/27
                  3s 98ms/step -
accuracy: 0.7826 - loss: 0.4872 - val accuracy: 0.7979 - val loss: 0.5635 -
learning_rate: 5.0000e-04
Epoch 7/20
27/27
                  3s 98ms/step -
accuracy: 0.7955 - loss: 0.4416 - val accuracy: 0.8085 - val loss: 0.3463 -
learning_rate: 5.0000e-04
Epoch 8/20
27/27
                  3s 97ms/step -
accuracy: 0.8053 - loss: 0.4130 - val_accuracy: 0.8298 - val_loss: 0.3492 -
learning_rate: 5.0000e-04
Epoch 9/20
                  3s 97ms/step -
27/27
accuracy: 0.8544 - loss: 0.3677 - val_accuracy: 0.8404 - val_loss: 0.3506 -
learning_rate: 5.0000e-04
Epoch 10/20
27/27
                  3s 97ms/step -
```

```
accuracy: 0.8611 - loss: 0.3603 - val_accuracy: 0.8617 - val_loss: 0.2943 -
learning_rate: 5.0000e-04
Epoch 11/20
27/27
                 3s 97ms/step -
accuracy: 0.8438 - loss: 0.3434 - val_accuracy: 0.8830 - val_loss: 0.3348 -
learning_rate: 2.5000e-04
Epoch 12/20
27/27
                 3s 98ms/step -
accuracy: 0.8594 - loss: 0.3544 - val_accuracy: 0.8723 - val_loss: 0.2727 -
learning_rate: 2.5000e-04
Epoch 13/20
27/27
                 3s 99ms/step -
accuracy: 0.8869 - loss: 0.2528 - val_accuracy: 0.8723 - val_loss: 0.3023 -
learning_rate: 2.5000e-04
Epoch 14/20
27/27
                 3s 98ms/step -
accuracy: 0.8975 - loss: 0.2524 - val_accuracy: 0.8511 - val_loss: 0.3737 -
learning_rate: 2.5000e-04
Epoch 15/20
27/27
                 3s 97ms/step -
accuracy: 0.8742 - loss: 0.2777 - val_accuracy: 0.8617 - val_loss: 0.3205 -
learning_rate: 2.5000e-04
Epoch 16/20
                 3s 100ms/step -
27/27
accuracy: 0.9109 - loss: 0.2373 - val_accuracy: 0.8830 - val_loss: 0.2283 -
learning_rate: 2.5000e-04
Epoch 17/20
27/27
                 Os 95ms/step -
accuracy: 0.9242 - loss: 0.2151Reached desired accuracy so cancelling training!
                 3s 98ms/step -
accuracy: 0.9233 - loss: 0.2161 - val_accuracy: 0.9149 - val_loss: 0.2582 -
learning_rate: 2.5000e-04
Current training accuracy: 0.8975265026092529
Current validation accuracy: 0.914893627166748
Reseting all weights...
Current number of trials: 3
```





WARNING:tensorflow:5 out of the last 7 calls to <function

TensorFlowTrainer.make\_predict\_function.<locals>.one\_step\_on\_data\_distributed at 0x17f228fe0> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3) passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2), @tf.function has reduce\_retracing=True option that can avoid unnecessary retracing. For (3), please refer to

https://www.tensorflow.org/guide/function#controlling\_retracing and https://www.tensorflow.org/api\_docs/python/tf/function for more details.

1/3 0s

47ms/stepWARNING:tensorflow:6 out of the last 9 calls to <function
TensorFlowTrainer.make\_predict\_function.<locals>.one\_step\_on\_data\_distributed at
0x17f228fe0> triggered tf.function retracing. Tracing is expensive and the
excessive number of tracings could be due to (1) creating @tf.function
repeatedly in a loop, (2) passing tensors with different shapes, (3) passing
Python objects instead of tensors. For (1), please define your @tf.function
outside of the loop. For (2), @tf.function has reduce\_retracing=True option that
can avoid unnecessary retracing. For (3), please refer to
https://www.tensorflow.org/guide/function#controlling\_retracing and
https://www.tensorflow.org/api\_docs/python/tf/function for more details.
3/3
0s 37ms/step

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,

support

'my\_model.keras')`.

precision recall f1-score

## Classification Report:

		Female	0.85	0.98	0.91	41
		Male	0.98	0.87	0.92	53
		accuracy			0.91	94
	m	acro avg	0.91	0.92	0.91	94
,	weig	hted avg	0.92	0.91	0.92	94
[1]:		accuracy	loss	val_accuracy	val_loss	learning_rate
	0	0.591284	0.700387	0.734043	0.632742	0.00050
	1	0.722026	0.578647	0.680851	0.568016	0.00050
	2	0.707892	0.555076	0.776596	0.483235	0.00050
	3	0.740872	0.512653	0.787234	0.534464	0.00050
	4	0.782097	0.474838	0.829787	0.429026	0.00050
	5	0.786808	0.470764	0.797872	0.563464	0.00050
	6	0.799764	0.429010	0.808511	0.346285	0.00050
	7	0.815077	0.410316	0.829787	0.349219	0.00050
	8	0.839812	0.380174	0.840426	0.350584	0.00050
	9	0.851590	0.364124	0.861702	0.294300	0.00050
	10	0.872792	0.296062	0.882979	0.334795	0.00025
	11	0.855124	0.327675	0.872340	0.272722	0.00025
	12	0.886926	0.270424	0.872340	0.302330	0.00025
	13	0.888104	0.276107	0.851064	0.373744	0.00025
	14	0.891637	0.257347	0.861702	0.320483	0.00025

```
[2]: from tensorflow.keras.models import Model
     from tensorflow.keras.utils import load_img, img_to_array
     img_size = img_size
     model = tf.keras.models.load_model("trained_model_run78_very_advanced_control.
      ⇔h5")
     # Load your personal image if you are interested to predict:
     \#your\_image\_path = "D: \ \ desktop files in Microsoft Studio_{\sqcup}
      \negLaptop\\Personal Photos\\Hossein_10.jpg"
     your_image_path = "/Users/hossein/Downloads/Hossein.png"
     img = load_img(your_image_path, target_size=(img_size, img_size))
     final_img = img_to_array(img)
     # Adding a batch dimension:
     final_img = np.expand_dims(final_img, axis=0)
     prediction = model.predict(final_img)
     result = "Female" if prediction > 0.5 else "Male"
     if result=="Female":
         confidence = (model.predict(final_img)[0][0])*100
     else:
         confidence = (1-model.predict(final_img)[0][0])*100
     print(f"Prediction result: {result} (confidence= {confidence:.2f} %)")
     # Visualize CNN Layers
     dummy_input = np.random.rand(1, img_size, img_size, 3) # Create random input_
      ⇔with correct shape
     model.predict(dummy_input) # Call the model to establish input tensors
     visualization_model = Model(inputs=model.inputs, outputs=[layer.output for_
      ⇒layer in model.layers])
     successive_feature_maps = visualization_model.predict(final_img)
     layer_names = [layer.name for layer in model.layers]
     for layer_name, feature_map in zip(layer_names, successive_feature_maps):
         if len(feature_map.shape) == 4: # Only visualize conv/maxpool layers
             n_features = feature_map.shape[-1] # Number of filters
             size = feature_map.shape[1] # Feature map size
             display_grid = np.zeros((size, size * n_features))
             for i in range(n_features):
                x = feature_map[0, :, :, i]
                 x -= x.mean()
                 x \neq (x.std() + 1e-8) # Normalize
                 x *= 64
                 x += 128
                 x = np.clip(x, 0, 255).astype('uint8') # Convert to image format
                 display_grid[:, i * size: (i + 1) * size] = x
```

```
scale = 20. / n_features
plt.figure(figsize=(scale * n_features, scale))
plt.title(layer_name)
plt.grid(False)
plt.imshow(display_grid, aspect='auto', cmap='cividis')
plt.show()
```

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 38ms/step 1/1 0s 16ms/step

Prediction result: Male (confidence= 91.61 %)

1/1 0s 15ms/step

/opt/anaconda3/envs/mytfenv/lib/python3.12/site-

packages/keras/src/models/functional.py:237: UserWarning: The structure of

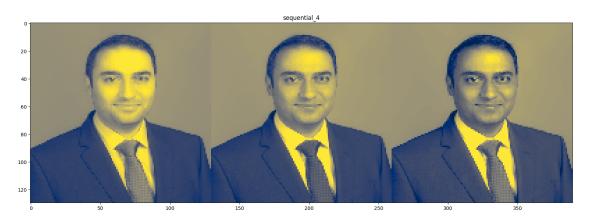
`inputs` doesn't match the expected structure.

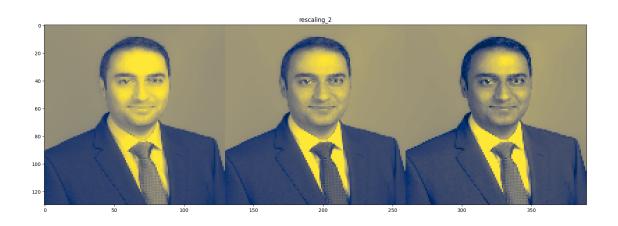
Expected: ['input\_layer\_5']

Received: inputs=Tensor(shape=(1, 130, 130, 3))

warnings.warn(msg)

1/1 0s 150ms/step



















				max_pooling2d_11			
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2 7	FO.	100	150	200	250	300	350
	30	100	130	200	230	300	330

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