faential 6classes vggface

June 13, 2023

```
[1]: !sudo apt-get update
     !sudo apt-get install -y gnupg2 curl
     !sudo curl -0 https://developer.download.nvidia.com/compute/cuda/repos/debian10/
     ⇒x86 64/cuda-ubuntu2004.pin
     !sudo mv cuda-ubuntu2004.pin /etc/apt/preferences.d/cuda-repository-pin-600
     !sudo curl -LO https://developer.download.nvidia.com/compute/cuda/11.4.2/
      ⇔local_installers/cuda-repo-debian10-11-4-local_11.4.2-470.57.02-1_amd64.deb
     !sudo dpkg -i cuda-repo-debian10-11-4-local_11.4.2-470.57.02-1_amd64.deb
     !sudo apt-get update
     !sudo apt-get -y install cuda
    Get:1 file:/var/cuda-repo-debian10-11-4-local InRelease
    Ign:1 file:/var/cuda-repo-debian10-11-4-local InRelease
    Get:2 file:/var/cuda-repo-debian10-11-4-local Release [564 B]
    Get:2 file:/var/cuda-repo-debian10-11-4-local Release [564 B]
    Get:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg [836 B]
    Get:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg [836 B]
    Hit:4 http://packages.cloud.google.com/apt gcsfuse-bullseye InRelease
    Hit:5 http://security.debian.org/debian-security bullseye-security InRelease
    Hit:6 http://deb.debian.org/debian bullseye InRelease
    Hit:7 http://packages.cloud.google.com/apt google-compute-engine-bullseye-stable
    InRelease
    Hit:8 http://deb.debian.org/debian bullseye-updates InRelease
    Hit:9 https://packages.cloud.google.com/apt google-fast-socket InRelease
    Hit:10 https://download.docker.com/linux/debian bullseye InRelease
    Hit:11 http://deb.debian.org/debian bullseye-backports InRelease
    Hit:12 http://packages.cloud.google.com/apt cloud-sdk-bullseye InRelease
    Hit:13 https://nvidia.github.io/libnvidia-container/stable/debian10/amd64
    InRelease
    Ign:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg
    Hit:14 https://nvidia.github.io/nvidia-container-runtime/stable/debian10/amd64
    InRelease
    Hit:15 https://nvidia.github.io/nvidia-docker/debian10/amd64 InRelease
    Hit:16 https://packages.cloud.google.com/apt kubernetes-xenial InRelease
    Reading package lists... Done
    W: GPG error: file:/var/cuda-repo-debian10-11-4-local Release: The following
    signatures couldn't be verified because the public key is not available:
    NO_PUBKEY F60F4B3D7FA2AF80
```

- E: The repository 'file:/var/cuda-repo-debian10-11-4-local Release' is not signed.
- N: Updating from such a repository can't be done securely, and is therefore disabled by default.
- N: See apt-secure(8) manpage for repository creation and user configuration details.

Reading package lists... Done

E: Unable to parse package file /etc/apt/preferences.d/cuda-repository-pin-600 (1)

Time Current % Total % Received % Xferd Average Speed Time Time Dload Upload Total Spent Left Speed 100 433 100 433 0 --:--:--0 436 % Total % Received % Xferd Average Speed Time Time Time Current Spent Dload Upload Total Left 100 2432M 100 2432M 187M 0 0:00:12 0:00:12 --:-- 199M (Reading database ... 128111 files and directories currently installed.) Preparing to unpack cuda-repo-debian10-11-4-local_11.4.2-470.57.02-1_amd64.deb

Unpacking cuda-repo-debian10-11-4-local (11.4.2-470.57.02-1) over (11.4.2-470.57.02-1) ...

Setting up cuda-repo-debian10-11-4-local (11.4.2-470.57.02-1) ...

The public CUDA GPG key does not appear to be installed.

To install the key, run this command:

sudo apt-key add /var/cuda-repo-debian10-11-4-local/7fa2af80.pub

- Get:1 file:/var/cuda-repo-debian10-11-4-local InRelease
- Ign:1 file:/var/cuda-repo-debian10-11-4-local InRelease
- Get:2 file:/var/cuda-repo-debian10-11-4-local Release [564 B]
- Get:2 file:/var/cuda-repo-debian10-11-4-local Release [564 B]
- Get:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg [836 B]
- Get:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg [836 B]
- Hit:4 http://deb.debian.org/debian bullseye InRelease
- Hit:5 http://packages.cloud.google.com/apt gcsfuse-bullseye InRelease
- Hit:6 http://security.debian.org/debian-security bullseye-security InRelease
- Hit:7 http://deb.debian.org/debian bullseye-updates InRelease
- Hit:8 http://deb.debian.org/debian bullseye-backports InRelease
- Hit:9 http://packages.cloud.google.com/apt google-compute-engine-bullseye-stable
 InRelease
- Hit:10 https://packages.cloud.google.com/apt google-fast-socket InRelease
- Hit:11 https://download.docker.com/linux/debian bullseye InRelease
- Hit:12 http://packages.cloud.google.com/apt cloud-sdk-bullseye InRelease
- Hit:13 https://nvidia.github.io/libnvidia-container/stable/debian10/amd64
 InRelease
- Ign:3 file:/var/cuda-repo-debian10-11-4-local Release.gpg
- Hit:14 https://nvidia.github.io/nvidia-container-runtime/stable/debian10/amd64
 InRelease
- Hit:15 https://nvidia.github.io/nvidia-docker/debian10/amd64 InRelease

Hit:16 https://packages.cloud.google.com/apt kubernetes-xenial InRelease
Reading package lists... Done

W: GPG error: file:/var/cuda-repo-debian10-11-4-local Release: The following signatures couldn't be verified because the public key is not available: NO PUBKEY F60F4B3D7FA2AF80

- E: The repository 'file:/var/cuda-repo-debian10-11-4-local Release' is not signed.
- N: Updating from such a repository can't be done securely, and is therefore disabled by default.
- N: See apt-secure(8) manpage for repository creation and user configuration details.

Reading package lists... Done

E: Unable to parse package file /etc/apt/preferences.d/cuda-repository-pin-600 (1)

[2]: !nvidia-smi

Tue Jun 13 07:39:29 2023

NVID	OIA-SMI	510.4	7.03	Driver	Version:	510.47.03	CUDA Versio	
GPU Fan	Name Temp	Perf	Persis Pwr:Us	tence-M age/Cap	Bus-Id 	Disp.A	Volatile GPU-Util 	Uncorr. ECC Compute M. MIG M.
O N/A 	Tesla 49C	T4 P0	26W	Off / 70W	0000000 0M:	0:00:04.0 Off iB / 15360MiB	 11% 	O Default N/A

[3]: | !pip install --user tensorflow==2.11.1

WARNING: Ignoring invalid distribution -eras

(/opt/conda/lib/python3.10/site-packages)

WARNING: Ignoring invalid distribution -rapt

(/opt/conda/lib/python3.10/site-packages)

Requirement already satisfied: tensorflow==2.11.1 in

/home/jupyter/.local/lib/python3.10/site-packages (2.11.1)

Requirement already satisfied: absl-py>=1.0.0 in /opt/conda/lib/python3.10/site-

```
packages (from tensorflow==2.11.1) (1.4.0)
Requirement already satisfied: astunparse>=1.6.0 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (1.6.3)
Requirement already satisfied: flatbuffers>=2.0 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (23.3.3)
Requirement already satisfied: gast<=0.4.0,>=0.2.1 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (0.4.0)
Requirement already satisfied: google-pasta>=0.1.1 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (0.2.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
(1.54.2)
Requirement already satisfied: h5py>=2.9.0 in /opt/conda/lib/python3.10/site-
packages (from tensorflow==2.11.1) (3.8.0)
Requirement already satisfied: keras<2.12,>=2.11.0 in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
(2.11.0)
Requirement already satisfied: libclang>=13.0.0 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (16.0.0)
Requirement already satisfied: numpy>=1.20 in /opt/conda/lib/python3.10/site-
packages (from tensorflow==2.11.1) (1.23.5)
Requirement already satisfied: opt-einsum>=2.3.2 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (3.3.0)
Requirement already satisfied: packaging in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
Requirement already satisfied: protobuf<3.20,>=3.9.2 in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
Requirement already satisfied: setuptools in /opt/conda/lib/python3.10/site-
packages (from tensorflow==2.11.1) (67.7.2)
Requirement already satisfied: six>=1.12.0 in /opt/conda/lib/python3.10/site-
packages (from tensorflow==2.11.1) (1.16.0)
Requirement already satisfied: tensorboard<2.12,>=2.11 in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
(2.11.2)
Requirement already satisfied: tensorflow-estimator<2.12,>=2.11.0 in
/home/jupyter/.local/lib/python3.10/site-packages (from tensorflow==2.11.1)
(2.11.0)
Requirement already satisfied: termcolor>=1.1.0 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (2.3.0)
Requirement already satisfied: typing-extensions>=3.6.6 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (4.5.0)
Requirement already satisfied: wrapt>=1.11.0 in /opt/conda/lib/python3.10/site-
packages (from tensorflow==2.11.1) (1.14.1)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
/opt/conda/lib/python3.10/site-packages (from tensorflow==2.11.1) (0.29.0)
Requirement already satisfied: wheel<1.0,>=0.23.0 in
```

```
/opt/conda/lib/python3.10/site-packages (from
astunparse>=1.6.0->tensorflow==2.11.1) (0.40.0)
Requirement already satisfied: google-auth<3,>=1.6.3 in
/opt/conda/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (2.17.3)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in
/home/jupyter/.local/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (0.4.6)
Requirement already satisfied: markdown>=2.6.8 in
/opt/conda/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (3.4.3)
Requirement already satisfied: requests<3,>=2.21.0 in
/opt/conda/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (2.28.2)
Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in
/home/jupyter/.local/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (0.6.1)
Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in
/opt/conda/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (1.8.1)
Requirement already satisfied: werkzeug>=1.0.1 in
/opt/conda/lib/python3.10/site-packages (from
tensorboard<2.12,>=2.11->tensorflow==2.11.1) (2.1.2)
Requirement already satisfied: pyparsing>=2.0.2 in
/opt/conda/lib/python3.10/site-packages (from packaging->tensorflow==2.11.1)
(3.0.9)
Requirement already satisfied: cachetools<6.0,>=2.0.0 in
/opt/conda/lib/python3.10/site-packages (from google-
auth<3,>=1.6.3->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (4.2.4)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
/opt/conda/lib/python3.10/site-packages (from google-
auth<3,>=1.6.3->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (0.2.7)
Requirement already satisfied: rsa<5,>=3.1.4 in /opt/conda/lib/python3.10/site-
packages (from google-
auth<3,>=1.6.3->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (4.9)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
/opt/conda/lib/python3.10/site-packages (from google-auth-
oauthlib<0.5,>=0.4.1->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (1.3.1)
Requirement already satisfied: charset-normalizer<4,>=2 in
/opt/conda/lib/python3.10/site-packages (from
requests<3,>=2.21.0->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (2.1.1)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/lib/python3.10/site-
packages (from requests<3,>=2.21.0->tensorboard<2.12,>=2.11->tensorflow==2.11.1)
(3.4)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in
/opt/conda/lib/python3.10/site-packages (from
requests<3,>=2.21.0->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (1.26.15)
Requirement already satisfied: certifi>=2017.4.17 in
```

```
/opt/conda/lib/python3.10/site-packages (from requests<3,>=2.21.0->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (2022.12.7) Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /opt/conda/lib/python3.10/site-packages (from pyasn1-modules>=0.2.1->google-auth<3,>=1.6.3->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (0.4.8) Requirement already satisfied: oauthlib>=3.0.0 in /opt/conda/lib/python3.10/site-packages (from requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.12,>=2.11->tensorflow==2.11.1) (3.2.2) WARNING: Ignoring invalid distribution -eras (/opt/conda/lib/python3.10/site-packages)

WARNING: Ignoring invalid distribution -rapt (/opt/conda/lib/python3.10/site-packages)
```

[4]: import tensorflow as tf

2023-06-13 07:39:33.186196: I tensorflow/core/platform/cpu_feature_guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 AVX512F FMA

To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

2023-06-13 07:39:34.227173: W tensorflow/compiler/xla/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'libnvinfer.so.7'; dlerror: libnvinfer.so.7: cannot open shared object file: No such file or directory; LD_LIBRARY_PATH: /usr/local/cuda/lib64:/usr/local/nccl2/lib:/usr/local/cuda/extras/CUPTI/lib64 2023-06-13 07:39:34.227292: W tensorflow/compiler/xla/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'libnvinfer_plugin.so.7'; dlerror: libnvinfer plugin.so.7: cannot open shared object file: No such file or

libnvinfer_plugin.so.7: cannot open shared object file: No such file or directory; LD_LIBRARY_PATH:

/usr/local/cuda/lib64:/usr/local/nccl2/lib:/usr/local/cuda/extras/CUPTI/lib64 2023-06-13 07:39:34.227302: W

tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Cannot dlopen some TensorRT libraries. If you would like to use Nvidia GPU with TensorRT, please make sure the missing libraries mentioned above are installed properly.

[5]: print(tf.__version__)

2.11.1

[6]: from keras.layers import Input, Lambda, Dense, Flatten from keras.models import Model

```
from keras.preprocessing import image
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Sequential
import numpy as np
from glob import glob
import matplotlib.pyplot as plt
```

```
[7]: import os import zipfile
```

define the directories containing your images

```
[8]: # variabel directory for training pict for each category
    train_oily_dir = os.path.join("/home/jupyter/content/faceSkin_tipe_train/oily")
    train_normal_dir = os.path.join("/home/jupyter/content/faceSkin_tipe_train/
      →normal")
    train_combination_dir =os.path.join("/home/jupyter/content/faceSkin_tipe_train/
      train_sensitive_dir = os.path.join("/home/jupyter/content/faceSkin_tipe_train/
      ⇔sensitive")
    train_dry_dir = os.path.join("/home/jupyter/content/faceSkin_tipe_train/dry")
    train_nonface_dir = os.path.join("/home/jupyter/content/faceSkin_tipe_train/
      ⇔nonface/")
     # variabel directory for validation pict for each category
    validation_oily_dir =os.path.join("/home/jupyter/content/

¬faceSkin_tipe_validation/oily")

    validation_normal_dir =os.path.join("/home/jupyter/content/
      →faceSkin_tipe_validation/normal")
    validation_combination_dir =os.path.join("/home/jupyter/content/
      →faceSkin_tipe_validation/combination")
    validation_sensitive_dir =os.path.join("/home/jupyter/content/

¬faceSkin tipe validation/sensitive")
    validation_dry_dir =os.path.join("/home/jupyter/content/
      →faceSkin_tipe_validation/dry")
    validation_nonface_dir =os.path.join("/home/jupyter/content/

¬faceSkin tipe validation/nonface")
```

filenames check in directory

```
[9]: train_dry_names = os.listdir(train_dry_dir)
    train_oily_names = os.listdir(train_oily_dir)
    train_normal_names = os.listdir(train_normal_dir)
    train_combination_names = os.listdir(train_combination_dir)
    train_sensitive_names = os.listdir(train_sensitive_dir)
    train_nonface_names = os.listdir(train_nonface_dir)
```

```
validation_dry_names = os.listdir(validation_dry_dir)
validation_oily_names = os.listdir(validation_oily_dir)
validation_normal_names = os.listdir(validation_normal_dir)
validation_combination_names = os.listdir(validation_combination_dir)
validation_sensitive_names = os.listdir(validation_sensitive_dir)
validation_nonface_names = os.listdir(validation_nonface_dir)
print(f'TRAIN SET DRY: {train_dry_names[:5]}')
print(f'TRAIN SET OILY: {train_dry_names[:5]}')
print(f'TRAIN SET SENSITIVE: {train_dry_names[:5]}')
print(f'TRAIN SET COMBINATION: {train_dry_names[:5]}')
print(f'TRAIN SET NORMAL: {train_dry_names[:5]}')
print(f'TRAIN SET NONFACE: {train_nonface_names[:5]} \n')
print(f'VALIDATION SET DRY: {validation_dry_names[:5]}')
print(f'VALIDATION SET OILY: {validation_dry_names[:5]}')
print(f'VALIDATION SET SENSITIVE: {validation_dry_names[:5]}')
print(f'VALIDATION SET COMBINATION: {validation_dry_names[:5]}')
print(f'VALIDATION SET NORMAL: {validation_dry_names[:5]}')
print(f'VALIDATION SET NONFACE: {validation_nonface_names[:5]}')
TRAIN SET DRY: ['f1-006-01.jpg', 'folder(185)3.jpg', 'folder(95)2.jpg',
'folder(224)2.jpg', 'folder(219)2.jpg']
TRAIN SET OILY: ['f1-006-01.jpg', 'folder(185)3.jpg', 'folder(95)2.jpg',
'folder(224)2.jpg', 'folder(219)2.jpg']
TRAIN SET SENSITIVE: ['f1-006-01.jpg', 'folder(185)3.jpg', 'folder(95)2.jpg',
'folder(224)2.jpg', 'folder(219)2.jpg']
TRAIN SET COMBINATION: ['f1-006-01.jpg', 'folder(185)3.jpg', 'folder(95)2.jpg',
'folder(224)2.jpg', 'folder(219)2.jpg']
TRAIN SET NORMAL: ['f1-006-01.jpg', 'folder(185)3.jpg', 'folder(95)2.jpg',
'folder(224)2.jpg', 'folder(219)2.jpg']
TRAIN SET NONFACE: ['000000013201.jpg', '000000017115.jpg', '000000002473.jpg',
'000000024021.jpg', '000000015335.jpg']
VALIDATION SET DRY: ['folder(7)2.jpg', 'dd.jpg', 'folder(72)3.jpg',
'folder(77)3.jpg', '1 (72).jpg']
VALIDATION SET OILY: ['folder(7)2.jpg', 'dd.jpg', 'folder(72)3.jpg',
'folder(77)3.jpg', '1 (72).jpg']
VALIDATION SET SENSITIVE: ['folder(7)2.jpg', 'dd.jpg', 'folder(72)3.jpg',
'folder(77)3.jpg', '1 (72).jpg']
VALIDATION SET COMBINATION: ['folder(7)2.jpg', 'dd.jpg', 'folder(72)3.jpg',
'folder(77)3.jpg', '1 (72).jpg']
VALIDATION SET NORMAL: ['folder(7)2.jpg', 'dd.jpg', 'folder(72)3.jpg',
'folder(77)3.jpg', '1 (72).jpg']
VALIDATION SET NONFACE: ['000000051008.jpg', '000000051976.jpg',
'000000045596.jpg', '000000047010.jpg', '000000045090.jpg']
```

chacking total number of images for each categories in training and validation directories

```
[10]: print(f'total training oily images: {len(os.listdir(train_oily_dir))}')
      print(f'total training dry images: {len(os.listdir(train_dry_dir))}')
      print(f'total training normal images: {len(os.listdir(train_normal_dir))}')
      print(f'total training combination images: {len(os.
       →listdir(train_combination_dir))}')
      print(f'total training sensitive images: {len(os.
       →listdir(train_sensitive_dir))}')
      print(f'total training nonface images: {len(os.listdir(train_nonface_dir))}\n')
      print(f'total validation oily images: {len(os.listdir(validation_oily_dir))}')
      print(f'total validation dry images: {len(os.listdir(validation_dry_dir))}')
      print(f'total validation normal images: {len(os.
       Glistdir(validation_normal_dir))}')
      print(f'total validation combination images: {len(os.
       ⇔listdir(validation_combination_dir))}')
      print(f'total validation sensitive images: {len(os.
       ⇔listdir(validation_sensitive_dir))}')
      print(f'total validation nonface images: {len(os.
       ⇔listdir(validation_nonface_dir))}')
     total training oily images: 209
     total training dry images: 207
     total training normal images: 99
     total training combination images: 110
     total training sensitive images: 202
     total training nonface images: 200
     total validation oily images: 53
     total validation dry images: 50
     total validation normal images: 25
     total validation combination images: 36
     total validation sensitive images: 49
     total validation nonface images: 50
 []:
```

1 Data preprocessing

using image data generator

```
[11]: #punya valiant
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os

# All images will be rescaled by 1./255
train_datagen = ImageDataGenerator(
```

```
rescale=1/255,
      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')
validation_datagen = ImageDataGenerator(rescale=1/255)
# Set the base directory where your data is located
base_directory = '/home/jupyter/content/'
# Get the list of subdirectories (classes) in the training directory
train_classes = [subdir for subdir in os.listdir(os.path.join(base_directory,_
d'faceSkin_tipe_train')) if os.path.isdir(os.path.join(base_directory,u)

¬'faceSkin_tipe_train', subdir))]
# Remove the ".ipynb_checkpoints" class from the list if it exists
if '.ipynb checkpoints' in train classes:
   train_classes.remove('.ipynb_checkpoints')
# Flow training images in batches of 128 using train_datagen generator
train_generator = train_datagen.flow_from_directory(
        os.path.join(base_directory, 'faceSkin_tipe_train'), # This is the_
 ⇔source directory for training images
       target_size=(224, 224), # All images will be resized to 224x224
       batch_size=32,
       class_mode='categorical',
       classes=train_classes)
# Get the list of subdirectories (classes) in the validation directory
validation_classes = [subdir for subdir in os.listdir(os.path.
 ⇔join(base_directory, 'faceSkin_tipe_validation')) if os.path.isdir(os.path.
 ⇒join(base_directory, 'faceSkin_tipe_validation', subdir))]
# Remove the ".ipynb checkpoints" class from the list if it exists
if '.ipynb_checkpoints' in validation_classes:
   validation_classes.remove('.ipynb_checkpoints')
# Flow validation images in batches of 128 using validation datagen generator
validation_generator = validation_datagen.flow_from_directory(
        os.path.join(base_directory, 'faceSkin_tipe_validation'), # This is_
→ the source directory for validation images
       target_size=(224, 224), # All images will be resized to 224x224
       batch_size=16,
```

```
class_mode='categorical',
              classes=validation_classes)
      # Get the class names from the generator's class_indices dictionary
      class_names = list(train_generator.class_indices.keys())
      # Print the class names
      print("Class Names:", class names)
     Found 1027 images belonging to 6 classes.
     Found 263 images belonging to 6 classes.
     Class Names: ['oily', 'normal', 'sensitive', 'dry', 'nonface', 'combination']
[12]: # Access the class indices
      class_indices = train_generator.class_indices
      # Print the list of classes
      print("List of Classes:")
      for class_name, class_index in class_indices.items():
          print(class_name, ":", class_index)
     List of Classes:
     oily : 0
     normal: 1
     sensitive: 2
     dry : 3
     nonface: 4
     combination: 5
     load pretrained model
[13]: | !python --version
     Python 3.10.10
[14]: !pip install keras_vggface
     WARNING: Ignoring invalid distribution -eras
     (/opt/conda/lib/python3.10/site-packages)
     WARNING: Ignoring invalid distribution -rapt
     (/opt/conda/lib/python3.10/site-packages)
     Requirement already satisfied: keras_vggface in
     /opt/conda/lib/python3.10/site-packages (0.6)
     Requirement already satisfied: numpy>=1.9.1 in /opt/conda/lib/python3.10/site-
     packages (from keras_vggface) (1.23.5)
     Requirement already satisfied: scipy>=0.14 in /opt/conda/lib/python3.10/site-
     packages (from keras_vggface) (1.9.3)
```

Requirement already satisfied: h5py in /opt/conda/lib/python3.10/site-packages (from keras_vggface) (3.8.0)
Requirement already satisfied: pillow in /opt/conda/lib/python3.10/site-packages (from keras_vggface) (9.5.0)
Requirement already satisfied: keras in /home/jupyter/.local/lib/python3.10/site-packages (from keras_vggface) (2.11.0)
Requirement already satisfied: six>=1.9.0 in /opt/conda/lib/python3.10/site-packages (from keras_vggface) (1.16.0)
Requirement already satisfied: pyyaml in /opt/conda/lib/python3.10/site-packages (from keras_vggface) (5.4.1)
WARNING: Ignoring invalid distribution -eras
(/opt/conda/lib/python3.10/site-packages)
WARNING: Ignoring invalid distribution -rapt
(/opt/conda/lib/python3.10/site-packages)

[15]: !pip install keras_applications

WARNING: Ignoring invalid distribution -eras

(/opt/conda/lib/python3.10/site-packages)

WARNING: Ignoring invalid distribution -rapt

(/opt/conda/lib/python3.10/site-packages)

Requirement already satisfied: keras_applications in

/opt/conda/lib/python3.10/site-packages (1.0.8)

Requirement already satisfied: numpy>=1.9.1 in /opt/conda/lib/python3.10/site-packages (from keras_applications) (1.23.5)

Requirement already satisfied: h5py in /opt/conda/lib/python3.10/site-packages (from keras_applications) (3.8.0)

WARNING: Ignoring invalid distribution -eras

(/opt/conda/lib/python3.10/site-packages)

WARNING: Ignoring invalid distribution -rapt

(/opt/conda/lib/python3.10/site-packages)

[16]: !pip show keras

WARNING: Ignoring invalid distribution -eras (/opt/conda/lib/python3.10/site-packages)
WARNING: Ignoring invalid distribution -rapt (/opt/conda/lib/python3.10/site-packages)

Name: keras

Version: 2.11.0

 ${\tt Summary:\ Deep\ learning\ for\ humans.}$

Home-page: https://keras.io/

Author: Keras team

Author-email: keras-users@googlegroups.com

License: Apache 2.0

Location: /home/jupyter/.local/lib/python3.10/site-packages

Requires:

Required-by: keras-vggface, tensorflow

[17]: import tensorflow.keras as keras

[18]: print(tf.__version__)

2.11.1

[20]: pre_trained_model.summary()

Model: "vggface_vgg16"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 224, 224, 3)]	0
conv1_1 (Conv2D)	(None, 224, 224, 64)	1792
conv1_2 (Conv2D)	(None, 224, 224, 64)	36928
pool1 (MaxPooling2D)	(None, 112, 112, 64)	0

```
conv2_1 (Conv2D)
                             (None, 112, 112, 128)
                                                       73856
conv2_2 (Conv2D)
                             (None, 112, 112, 128)
                                                       147584
pool2 (MaxPooling2D)
                             (None, 56, 56, 128)
conv3_1 (Conv2D)
                             (None, 56, 56, 256)
                                                       295168
conv3_2 (Conv2D)
                             (None, 56, 56, 256)
                                                       590080
conv3_3 (Conv2D)
                             (None, 56, 56, 256)
                                                       590080
pool3 (MaxPooling2D)
                             (None, 28, 28, 256)
conv4_1 (Conv2D)
                             (None, 28, 28, 512)
                                                       1180160
conv4_2 (Conv2D)
                             (None, 28, 28, 512)
                                                       2359808
                             (None, 28, 28, 512)
conv4 3 (Conv2D)
                                                       2359808
pool4 (MaxPooling2D)
                             (None, 14, 14, 512)
                             (None, 14, 14, 512)
conv5_1 (Conv2D)
                                                       2359808
conv5_2 (Conv2D)
                             (None, 14, 14, 512)
                                                       2359808
                             (None, 14, 14, 512)
conv5_3 (Conv2D)
                                                       2359808
pool5 (MaxPooling2D)
                             (None, 7, 7, 512)
```

Total params: 14,714,688 Trainable params: 0

Trainable params. 0

Non-trainable params: 14,714,688

```
our_model = tf.keras.models.Sequential(pre_trained_model)
our_model.add(tf.keras.layers.Conv2D(filters = 1024, kernel_size = (3,3),
→activation = "relu", padding = "same"))
our_model.add(tf.keras.layers.Conv2D(filters = 512, kernel_size = (3,3),
→activation = "relu", padding = "same"))
```

```
[22]: our_model.summary()
```

Model: "sequential"

Layer (type) Output Shape Param #

```
vggface_vgg16 (Functional) (None, 7, 7, 512)
                                                         14714688
      conv2d (Conv2D)
                                 (None, 7, 7, 1024) 4719616
      conv2d 1 (Conv2D)
                                 (None, 7, 7, 512)
     Total params: 24,153,408
     Trainable params: 9,438,720
     Non-trainable params: 14,714,688
[24]: # Choose `mixed 7` as the last layer of your base model
     last_layer = our_model.get_layer('conv2d')
     print('last layer output shape: ', last_layer.output_shape)
     last_output = last_layer.output
     print('last layer output: ', last_output)
     last layer output shape: (None, 7, 7, 1024)
     last layer output: KerasTensor(type_spec=TensorSpec(shape=(None, 7, 7, 1024),
     dtype=tf.float32, name=None), name='conv2d/Relu:0', description="created by
     layer 'conv2d'")
        add dense layer depends on the classification (5 category)
[25]: from tensorflow.keras.optimizers import RMSprop
     from tensorflow.keras import Model
     # Flatten the output layer to 1 dimension
     x = layers.Flatten()(last_output)
      # Add a fully connected layer with 1,024 hidden units and ReLU activation
     x = layers.Dense(1024, activation="ReLU")(x)
     # Add a dropout rate of 0.2
```

```
# Print the model summary. See your dense network connected at the end.
model.summary()

Model: "model"

Layer (type) Output Shape Param #
```

x = layers.Dropout(0.2)(x)

model = Model(our_model.input, x)

Add a final sigmoid layer for classification
x = layers.Dense(6, activation='softmax')(x)

Append the dense network to the base model

```
vggface_vgg16_input (InputL [(None, 224, 224, 3)]
      ayer)
      vggface_vgg16 (Functional)
                                 (None, 7, 7, 512)
                                                          14714688
      conv2d (Conv2D)
                                 (None, 7, 7, 1024)
                                                          4719616
      flatten (Flatten)
                                 (None, 50176)
      dense (Dense)
                                 (None, 1024)
                                                          51381248
                                 (None, 1024)
      dropout (Dropout)
      dense_1 (Dense)
                                 (None, 6)
                                                           6150
     ______
     Total params: 70,821,702
     Trainable params: 56,107,014
     Non-trainable params: 14,714,688
     compiling the model
[26]: from tensorflow.keras.optimizers import Adam
     model.compile(loss=tf.keras.losses.CategoricalCrossentropy(),
                   optimizer=tf.keras.optimizers.Adam(learning_rate = 0.0001),
                   metrics=['accuracy'])
      #membuat callback untuk menentukan learning rate terbaik
      \#lr\ scheduler = tf.keras.callbacks.LearningRateScheduler(lambda\ epoch:1e-4\ *\ 10_{\sqcup}
       →**(epoch/20))
     Training
[27]: len(validation_generator)
[27]: 17
[28]: history = model.fit(
           train_generator,
           steps_per_epoch=len(train_generator),
           epochs=150,
           \#callbacks = [lr_scheduler],
           verbose=2,
           validation_data = validation_generator,
           validation_steps=len(validation_generator))
```

Epoch 1/150

```
33/33 - 81s - loss: 1.4597 - accuracy: 0.4070 - val_loss: 1.2538 - val_accuracy:
0.4791 - 81s/epoch - 2s/step
Epoch 2/150
33/33 - 65s - loss: 1.1511 - accuracy: 0.5209 - val_loss: 1.1242 - val_accuracy:
0.4943 - 65s/epoch - 2s/step
Epoch 3/150
33/33 - 65s - loss: 1.1032 - accuracy: 0.5482 - val_loss: 1.1461 - val_accuracy:
0.5019 - 65s/epoch - 2s/step
Epoch 4/150
33/33 - 64s - loss: 1.0574 - accuracy: 0.5463 - val_loss: 1.1241 - val_accuracy:
0.5323 - 64s/epoch - 2s/step
Epoch 5/150
33/33 - 64s - loss: 1.0179 - accuracy: 0.5618 - val_loss: 1.0852 - val_accuracy:
0.5627 - 64s/epoch - 2s/step
Epoch 6/150
33/33 - 64s - loss: 0.9656 - accuracy: 0.5988 - val_loss: 1.0663 - val_accuracy:
0.5551 - 64s/epoch - 2s/step
Epoch 7/150
33/33 - 64s - loss: 0.9088 - accuracy: 0.6280 - val_loss: 1.0729 - val_accuracy:
0.5779 - 64s/epoch - 2s/step
Epoch 8/150
33/33 - 65s - loss: 0.9173 - accuracy: 0.6144 - val_loss: 1.0203 - val_accuracy:
0.5894 - 65s/epoch - 2s/step
Epoch 9/150
33/33 - 64s - loss: 0.8500 - accuracy: 0.6650 - val_loss: 1.0865 - val_accuracy:
0.5932 - 64s/epoch - 2s/step
Epoch 10/150
33/33 - 65s - loss: 0.9249 - accuracy: 0.6232 - val_loss: 0.9578 - val_accuracy:
0.5856 - 65s/epoch - 2s/step
Epoch 11/150
33/33 - 64s - loss: 0.8237 - accuracy: 0.6670 - val_loss: 1.0333 - val_accuracy:
0.6198 - 64s/epoch - 2s/step
Epoch 12/150
33/33 - 64s - loss: 0.7877 - accuracy: 0.6913 - val_loss: 1.0366 - val_accuracy:
0.6008 - 64s/epoch - 2s/step
Epoch 13/150
33/33 - 64s - loss: 0.7939 - accuracy: 0.6943 - val loss: 0.9870 - val accuracy:
0.6046 - 64s/epoch - 2s/step
Epoch 14/150
33/33 - 64s - loss: 0.7415 - accuracy: 0.7225 - val_loss: 1.0818 - val_accuracy:
0.6236 - 64s/epoch - 2s/step
Epoch 15/150
33/33 - 64s - loss: 0.7110 - accuracy: 0.7371 - val_loss: 0.9597 - val_accuracy:
0.5970 - 64s/epoch - 2s/step
Epoch 16/150
33/33 - 64s - loss: 0.6974 - accuracy: 0.7352 - val_loss: 0.9505 - val_accuracy:
0.6046 - 64s/epoch - 2s/step
Epoch 17/150
```

```
33/33 - 64s - loss: 0.6776 - accuracy: 0.7342 - val_loss: 1.2675 - val_accuracy:
0.5703 - 64s/epoch - 2s/step
Epoch 18/150
33/33 - 64s - loss: 0.6460 - accuracy: 0.7400 - val_loss: 0.9887 - val_accuracy:
0.6502 - 64s/epoch - 2s/step
Epoch 19/150
33/33 - 65s - loss: 0.6533 - accuracy: 0.7556 - val_loss: 0.9211 - val_accuracy:
0.6426 - 65s/epoch - 2s/step
Epoch 20/150
33/33 - 64s - loss: 0.6313 - accuracy: 0.7614 - val_loss: 0.9412 - val_accuracy:
0.6312 - 64s/epoch - 2s/step
Epoch 21/150
33/33 - 64s - loss: 0.5919 - accuracy: 0.7712 - val_loss: 0.9678 - val_accuracy:
0.6578 - 64s/epoch - 2s/step
Epoch 22/150
33/33 - 64s - loss: 0.5753 - accuracy: 0.7780 - val_loss: 0.9066 - val_accuracy:
0.6616 - 64s/epoch - 2s/step
Epoch 23/150
33/33 - 64s - loss: 0.5637 - accuracy: 0.7936 - val_loss: 0.9728 - val_accuracy:
0.6388 - 64s/epoch - 2s/step
Epoch 24/150
33/33 - 64s - loss: 0.5439 - accuracy: 0.8101 - val_loss: 0.9285 - val_accuracy:
0.6730 - 64s/epoch - 2s/step
Epoch 25/150
33/33 - 64s - loss: 0.5408 - accuracy: 0.7916 - val_loss: 0.9882 - val_accuracy:
0.6350 - 64s/epoch - 2s/step
Epoch 26/150
33/33 - 64s - loss: 0.4968 - accuracy: 0.8179 - val_loss: 0.9303 - val_accuracy:
0.6730 - 64s/epoch - 2s/step
Epoch 27/150
33/33 - 63s - loss: 0.4973 - accuracy: 0.8257 - val_loss: 0.9369 - val_accuracy:
0.6844 - 63s/epoch - 2s/step
Epoch 28/150
33/33 - 64s - loss: 0.5158 - accuracy: 0.8023 - val_loss: 0.9758 - val_accuracy:
0.6464 - 64s/epoch - 2s/step
Epoch 29/150
33/33 - 64s - loss: 0.4647 - accuracy: 0.8335 - val loss: 0.8932 - val accuracy:
0.6844 - 64s/epoch - 2s/step
Epoch 30/150
33/33 - 64s - loss: 0.4388 - accuracy: 0.8315 - val_loss: 0.9864 - val_accuracy:
0.6806 - 64s/epoch - 2s/step
Epoch 31/150
33/33 - 64s - loss: 0.4326 - accuracy: 0.8423 - val_loss: 0.9210 - val_accuracy:
0.6692 - 64s/epoch - 2s/step
Epoch 32/150
33/33 - 64s - loss: 0.3967 - accuracy: 0.8578 - val_loss: 1.0397 - val_accuracy:
0.6654 - 64s/epoch - 2s/step
Epoch 33/150
```

```
33/33 - 63s - loss: 0.3639 - accuracy: 0.8773 - val_loss: 0.9850 - val_accuracy:
0.6692 - 63s/epoch - 2s/step
Epoch 34/150
33/33 - 64s - loss: 0.3861 - accuracy: 0.8685 - val_loss: 1.1108 - val_accuracy:
0.6502 - 64s/epoch - 2s/step
Epoch 35/150
33/33 - 65s - loss: 0.3859 - accuracy: 0.8442 - val loss: 1.0766 - val accuracy:
0.6730 - 65s/epoch - 2s/step
Epoch 36/150
33/33 - 64s - loss: 0.3841 - accuracy: 0.8588 - val_loss: 0.9733 - val_accuracy:
0.6730 - 64s/epoch - 2s/step
Epoch 37/150
33/33 - 64s - loss: 0.3447 - accuracy: 0.8676 - val_loss: 0.9092 - val_accuracy:
0.6882 - 64s/epoch - 2s/step
Epoch 38/150
33/33 - 64s - loss: 0.3243 - accuracy: 0.8929 - val_loss: 0.9704 - val_accuracy:
0.6844 - 64s/epoch - 2s/step
Epoch 39/150
33/33 - 64s - loss: 0.3053 - accuracy: 0.8890 - val_loss: 0.9947 - val_accuracy:
0.6920 - 64s/epoch - 2s/step
Epoch 40/150
33/33 - 64s - loss: 0.3098 - accuracy: 0.8948 - val_loss: 0.9600 - val_accuracy:
0.7034 - 64s/epoch - 2s/step
Epoch 41/150
33/33 - 64s - loss: 0.2995 - accuracy: 0.8997 - val_loss: 1.0062 - val_accuracy:
0.6616 - 64s/epoch - 2s/step
Epoch 42/150
33/33 - 65s - loss: 0.3067 - accuracy: 0.8919 - val_loss: 1.0887 - val_accuracy:
0.6768 - 65s/epoch - 2s/step
Epoch 43/150
33/33 - 63s - loss: 0.2986 - accuracy: 0.8958 - val_loss: 1.0407 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 44/150
33/33 - 63s - loss: 0.2972 - accuracy: 0.8929 - val_loss: 0.9825 - val_accuracy:
0.6920 - 63s/epoch - 2s/step
Epoch 45/150
33/33 - 64s - loss: 0.2785 - accuracy: 0.9026 - val loss: 1.0035 - val accuracy:
0.7110 - 64s/epoch - 2s/step
Epoch 46/150
33/33 - 64s - loss: 0.2591 - accuracy: 0.9143 - val_loss: 1.0179 - val_accuracy:
0.7224 - 64s/epoch - 2s/step
Epoch 47/150
33/33 - 63s - loss: 0.2304 - accuracy: 0.9231 - val_loss: 1.2240 - val_accuracy:
0.6806 - 63s/epoch - 2s/step
Epoch 48/150
33/33 - 63s - loss: 0.2621 - accuracy: 0.9007 - val_loss: 1.0384 - val_accuracy:
0.7110 - 63s/epoch - 2s/step
Epoch 49/150
```

```
33/33 - 63s - loss: 0.2274 - accuracy: 0.9260 - val_loss: 0.9883 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 50/150
33/33 - 63s - loss: 0.2141 - accuracy: 0.9328 - val_loss: 1.1862 - val_accuracy:
0.6996 - 63s/epoch - 2s/step
Epoch 51/150
33/33 - 64s - loss: 0.2161 - accuracy: 0.9318 - val loss: 1.0779 - val accuracy:
0.7148 - 64s/epoch - 2s/step
Epoch 52/150
33/33 - 65s - loss: 0.2198 - accuracy: 0.9202 - val_loss: 1.0861 - val_accuracy:
0.6996 - 65s/epoch - 2s/step
Epoch 53/150
33/33 - 66s - loss: 0.1967 - accuracy: 0.9406 - val_loss: 1.2105 - val_accuracy:
0.6806 - 66s/epoch - 2s/step
Epoch 54/150
33/33 - 63s - loss: 0.1987 - accuracy: 0.9260 - val_loss: 1.2019 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 55/150
33/33 - 63s - loss: 0.2029 - accuracy: 0.9241 - val_loss: 1.1999 - val_accuracy:
0.6882 - 63s/epoch - 2s/step
Epoch 56/150
33/33 - 63s - loss: 0.2097 - accuracy: 0.9289 - val_loss: 1.1759 - val_accuracy:
0.6882 - 63s/epoch - 2s/step
Epoch 57/150
33/33 - 63s - loss: 0.2163 - accuracy: 0.9309 - val_loss: 1.0700 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 58/150
33/33 - 63s - loss: 0.2284 - accuracy: 0.9221 - val_loss: 1.3474 - val_accuracy:
0.6730 - 63s/epoch - 2s/step
Epoch 59/150
33/33 - 63s - loss: 0.2007 - accuracy: 0.9309 - val_loss: 1.2759 - val_accuracy:
0.6920 - 63s/epoch - 2s/step
Epoch 60/150
33/33 - 63s - loss: 0.2017 - accuracy: 0.9299 - val_loss: 1.0278 - val_accuracy:
0.7262 - 63s/epoch - 2s/step
Epoch 61/150
33/33 - 64s - loss: 0.1594 - accuracy: 0.9513 - val loss: 1.1543 - val accuracy:
0.7148 - 64s/epoch - 2s/step
Epoch 62/150
33/33 - 63s - loss: 0.1746 - accuracy: 0.9455 - val_loss: 1.1713 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 63/150
33/33 - 63s - loss: 0.1435 - accuracy: 0.9591 - val_loss: 1.2482 - val_accuracy:
0.7110 - 63s/epoch - 2s/step
Epoch 64/150
33/33 - 63s - loss: 0.1432 - accuracy: 0.9533 - val_loss: 1.1989 - val_accuracy:
0.7300 - 63s/epoch - 2s/step
Epoch 65/150
```

```
33/33 - 63s - loss: 0.1592 - accuracy: 0.9416 - val_loss: 1.1847 - val_accuracy:
0.7110 - 63s/epoch - 2s/step
Epoch 66/150
33/33 - 63s - loss: 0.1498 - accuracy: 0.9523 - val_loss: 1.1832 - val_accuracy:
0.7452 - 63s/epoch - 2s/step
Epoch 67/150
33/33 - 63s - loss: 0.1871 - accuracy: 0.9435 - val loss: 1.2513 - val accuracy:
0.6920 - 63s/epoch - 2s/step
Epoch 68/150
33/33 - 63s - loss: 0.1530 - accuracy: 0.9503 - val_loss: 1.2773 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 69/150
33/33 - 63s - loss: 0.1380 - accuracy: 0.9620 - val_loss: 1.2051 - val_accuracy:
0.7110 - 63s/epoch - 2s/step
Epoch 70/150
33/33 - 63s - loss: 0.1345 - accuracy: 0.9494 - val_loss: 1.2321 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 71/150
33/33 - 64s - loss: 0.1341 - accuracy: 0.9601 - val_loss: 1.3268 - val_accuracy:
0.7034 - 64s/epoch - 2s/step
Epoch 72/150
33/33 - 64s - loss: 0.1799 - accuracy: 0.9445 - val_loss: 1.1474 - val_accuracy:
0.7110 - 64s/epoch - 2s/step
Epoch 73/150
33/33 - 63s - loss: 0.1947 - accuracy: 0.9396 - val_loss: 1.1353 - val_accuracy:
0.7110 - 63s/epoch - 2s/step
Epoch 74/150
33/33 - 63s - loss: 0.1356 - accuracy: 0.9581 - val_loss: 1.2056 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 75/150
33/33 - 63s - loss: 0.1755 - accuracy: 0.9357 - val_loss: 1.1835 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 76/150
33/33 - 63s - loss: 0.1171 - accuracy: 0.9620 - val_loss: 1.1525 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 77/150
33/33 - 63s - loss: 0.1099 - accuracy: 0.9649 - val loss: 1.2640 - val accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 78/150
33/33 - 64s - loss: 0.1257 - accuracy: 0.9581 - val_loss: 1.2071 - val_accuracy:
0.7414 - 64s/epoch - 2s/step
Epoch 79/150
33/33 - 63s - loss: 0.1618 - accuracy: 0.9503 - val_loss: 1.2430 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 80/150
33/33 - 63s - loss: 0.2022 - accuracy: 0.9211 - val_loss: 1.2626 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 81/150
```

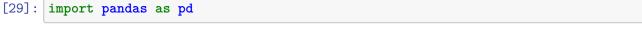
```
33/33 - 63s - loss: 0.1254 - accuracy: 0.9562 - val_loss: 1.3748 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 82/150
33/33 - 64s - loss: 0.1384 - accuracy: 0.9503 - val_loss: 1.2462 - val_accuracy:
0.7338 - 64s/epoch - 2s/step
Epoch 83/150
33/33 - 63s - loss: 0.1052 - accuracy: 0.9698 - val loss: 1.3474 - val accuracy:
0.7414 - 63s/epoch - 2s/step
Epoch 84/150
33/33 - 63s - loss: 0.1121 - accuracy: 0.9640 - val_loss: 1.3500 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 85/150
33/33 - 63s - loss: 0.1222 - accuracy: 0.9572 - val_loss: 1.3622 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 86/150
33/33 - 63s - loss: 0.1560 - accuracy: 0.9503 - val_loss: 1.3202 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 87/150
33/33 - 63s - loss: 0.1192 - accuracy: 0.9620 - val_loss: 1.3695 - val_accuracy:
0.6958 - 63s/epoch - 2s/step
Epoch 88/150
33/33 - 62s - loss: 0.0896 - accuracy: 0.9718 - val_loss: 1.3514 - val_accuracy:
0.7148 - 62s/epoch - 2s/step
Epoch 89/150
33/33 - 63s - loss: 0.0830 - accuracy: 0.9737 - val_loss: 1.3805 - val_accuracy:
0.7148 - 63s/epoch - 2s/step
Epoch 90/150
33/33 - 63s - loss: 0.2444 - accuracy: 0.9309 - val_loss: 1.1691 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 91/150
33/33 - 63s - loss: 0.1140 - accuracy: 0.9649 - val_loss: 1.1260 - val_accuracy:
0.7414 - 63s/epoch - 2s/step
Epoch 92/150
33/33 - 63s - loss: 0.1031 - accuracy: 0.9708 - val_loss: 1.2285 - val_accuracy:
0.7490 - 63s/epoch - 2s/step
Epoch 93/150
33/33 - 63s - loss: 0.0928 - accuracy: 0.9747 - val loss: 1.3104 - val accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 94/150
33/33 - 63s - loss: 0.1029 - accuracy: 0.9630 - val_loss: 1.2610 - val_accuracy:
0.7414 - 63s/epoch - 2s/step
Epoch 95/150
33/33 - 63s - loss: 0.0953 - accuracy: 0.9669 - val_loss: 1.3800 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 96/150
33/33 - 64s - loss: 0.1021 - accuracy: 0.9659 - val_loss: 1.2352 - val_accuracy:
0.7376 - 64s/epoch - 2s/step
Epoch 97/150
```

```
33/33 - 64s - loss: 0.0929 - accuracy: 0.9659 - val_loss: 1.3739 - val_accuracy:
0.7376 - 64s/epoch - 2s/step
Epoch 98/150
33/33 - 63s - loss: 0.1324 - accuracy: 0.9533 - val_loss: 1.3296 - val_accuracy:
0.7567 - 63s/epoch - 2s/step
Epoch 99/150
33/33 - 63s - loss: 0.0909 - accuracy: 0.9698 - val loss: 1.2161 - val accuracy:
0.7567 - 63s/epoch - 2s/step
Epoch 100/150
33/33 - 64s - loss: 0.1678 - accuracy: 0.9484 - val_loss: 1.3049 - val_accuracy:
0.7376 - 64s/epoch - 2s/step
Epoch 101/150
33/33 - 63s - loss: 0.0882 - accuracy: 0.9766 - val_loss: 1.3236 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 102/150
33/33 - 63s - loss: 0.0976 - accuracy: 0.9688 - val_loss: 1.3043 - val_accuracy:
0.7490 - 63s/epoch - 2s/step
Epoch 103/150
33/33 - 63s - loss: 0.1124 - accuracy: 0.9669 - val_loss: 1.4611 - val_accuracy:
0.7148 - 63s/epoch - 2s/step
Epoch 104/150
33/33 - 63s - loss: 0.0941 - accuracy: 0.9708 - val_loss: 1.6989 - val_accuracy:
0.6806 - 63s/epoch - 2s/step
Epoch 105/150
33/33 - 63s - loss: 0.1423 - accuracy: 0.9513 - val_loss: 1.2583 - val_accuracy:
0.7719 - 63s/epoch - 2s/step
Epoch 106/150
33/33 - 63s - loss: 0.1664 - accuracy: 0.9484 - val_loss: 1.2530 - val_accuracy:
0.7757 - 63s/epoch - 2s/step
Epoch 107/150
33/33 - 63s - loss: 0.1005 - accuracy: 0.9708 - val_loss: 1.3407 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 108/150
33/33 - 64s - loss: 0.0726 - accuracy: 0.9796 - val_loss: 1.2361 - val_accuracy:
0.7414 - 64s/epoch - 2s/step
Epoch 109/150
33/33 - 63s - loss: 0.1003 - accuracy: 0.9679 - val loss: 1.3945 - val accuracy:
0.7414 - 63s/epoch - 2s/step
Epoch 110/150
33/33 - 63s - loss: 0.0776 - accuracy: 0.9757 - val_loss: 1.4965 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 111/150
33/33 - 63s - loss: 0.0897 - accuracy: 0.9727 - val_loss: 1.4180 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 112/150
33/33 - 63s - loss: 0.0731 - accuracy: 0.9757 - val_loss: 1.4165 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 113/150
```

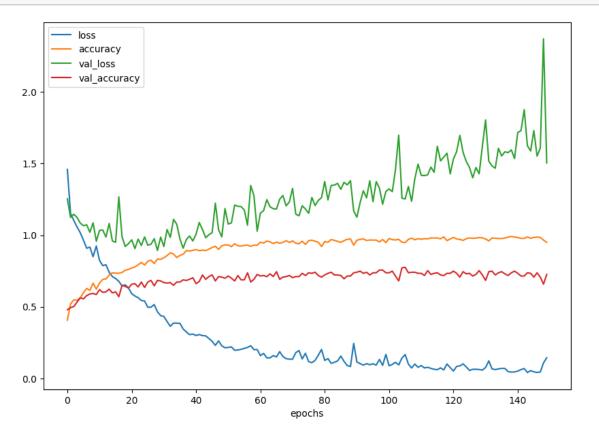
```
33/33 - 63s - loss: 0.0774 - accuracy: 0.9747 - val_loss: 1.4216 - val_accuracy:
0.7529 - 63s/epoch - 2s/step
Epoch 114/150
33/33 - 63s - loss: 0.0701 - accuracy: 0.9805 - val_loss: 1.4757 - val_accuracy:
0.7262 - 63s/epoch - 2s/step
Epoch 115/150
33/33 - 63s - loss: 0.0636 - accuracy: 0.9805 - val loss: 1.4392 - val accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 116/150
33/33 - 63s - loss: 0.0609 - accuracy: 0.9805 - val_loss: 1.6212 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 117/150
33/33 - 63s - loss: 0.0736 - accuracy: 0.9747 - val_loss: 1.5188 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 118/150
33/33 - 63s - loss: 0.0592 - accuracy: 0.9883 - val_loss: 1.5453 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 119/150
33/33 - 63s - loss: 0.1006 - accuracy: 0.9620 - val_loss: 1.5726 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 120/150
33/33 - 63s - loss: 0.0755 - accuracy: 0.9737 - val_loss: 1.4286 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 121/150
33/33 - 63s - loss: 0.0513 - accuracy: 0.9844 - val_loss: 1.5320 - val_accuracy:
0.7490 - 63s/epoch - 2s/step
Epoch 122/150
33/33 - 63s - loss: 0.0838 - accuracy: 0.9737 - val_loss: 1.5842 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 123/150
33/33 - 63s - loss: 0.0873 - accuracy: 0.9708 - val_loss: 1.6973 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 124/150
33/33 - 63s - loss: 0.1014 - accuracy: 0.9640 - val_loss: 1.5791 - val_accuracy:
0.7452 - 63s/epoch - 2s/step
Epoch 125/150
33/33 - 63s - loss: 0.0799 - accuracy: 0.9757 - val loss: 1.5145 - val accuracy:
0.7300 - 63s/epoch - 2s/step
Epoch 126/150
33/33 - 63s - loss: 0.0557 - accuracy: 0.9805 - val_loss: 1.4733 - val_accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 127/150
33/33 - 63s - loss: 0.0634 - accuracy: 0.9786 - val_loss: 1.4010 - val_accuracy:
0.7148 - 63s/epoch - 2s/step
Epoch 128/150
33/33 - 63s - loss: 0.0639 - accuracy: 0.9796 - val_loss: 1.4730 - val_accuracy:
0.7262 - 63s/epoch - 2s/step
Epoch 129/150
```

```
33/33 - 64s - loss: 0.0621 - accuracy: 0.9834 - val_loss: 1.4287 - val_accuracy:
0.7529 - 64s/epoch - 2s/step
Epoch 130/150
33/33 - 63s - loss: 0.0577 - accuracy: 0.9815 - val_loss: 1.6208 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 131/150
33/33 - 63s - loss: 0.0763 - accuracy: 0.9737 - val_loss: 1.8050 - val_accuracy:
0.6844 - 63s/epoch - 2s/step
Epoch 132/150
33/33 - 63s - loss: 0.1223 - accuracy: 0.9601 - val_loss: 1.5171 - val_accuracy:
0.7452 - 63s/epoch - 2s/step
Epoch 133/150
33/33 - 63s - loss: 0.0664 - accuracy: 0.9805 - val_loss: 1.4838 - val_accuracy:
0.7490 - 63s/epoch - 2s/step
Epoch 134/150
33/33 - 63s - loss: 0.0613 - accuracy: 0.9786 - val_loss: 1.4680 - val_accuracy:
0.7224 - 63s/epoch - 2s/step
Epoch 135/150
33/33 - 64s - loss: 0.0664 - accuracy: 0.9776 - val_loss: 1.6072 - val_accuracy:
0.7376 - 64s/epoch - 2s/step
Epoch 136/150
33/33 - 63s - loss: 0.0699 - accuracy: 0.9766 - val_loss: 1.5532 - val_accuracy:
0.7452 - 63s/epoch - 2s/step
Epoch 137/150
33/33 - 64s - loss: 0.0695 - accuracy: 0.9796 - val_loss: 1.5827 - val_accuracy:
0.7300 - 64s/epoch - 2s/step
Epoch 138/150
33/33 - 63s - loss: 0.0475 - accuracy: 0.9873 - val_loss: 1.5768 - val_accuracy:
0.7186 - 63s/epoch - 2s/step
Epoch 139/150
33/33 - 63s - loss: 0.0438 - accuracy: 0.9912 - val_loss: 1.5955 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 140/150
33/33 - 64s - loss: 0.0452 - accuracy: 0.9873 - val_loss: 1.5358 - val_accuracy:
0.7490 - 64s/epoch - 2s/step
Epoch 141/150
33/33 - 63s - loss: 0.0508 - accuracy: 0.9834 - val loss: 1.7170 - val accuracy:
0.7338 - 63s/epoch - 2s/step
Epoch 142/150
33/33 - 63s - loss: 0.0616 - accuracy: 0.9776 - val_loss: 1.7303 - val_accuracy:
0.7148 - 63s/epoch - 2s/step
Epoch 143/150
33/33 - 63s - loss: 0.0689 - accuracy: 0.9776 - val_loss: 1.8764 - val_accuracy:
0.7148 - 63s/epoch - 2s/step
Epoch 144/150
33/33 - 63s - loss: 0.0410 - accuracy: 0.9883 - val_loss: 1.6247 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 145/150
```

```
33/33 - 62s - loss: 0.0546 - accuracy: 0.9786 - val_loss: 1.5878 - val_accuracy:
0.7338 - 62s/epoch - 2s/step
Epoch 146/150
33/33 - 63s - loss: 0.0459 - accuracy: 0.9844 - val_loss: 1.7303 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 147/150
33/33 - 63s - loss: 0.0419 - accuracy: 0.9873 - val_loss: 1.5522 - val_accuracy:
0.7376 - 63s/epoch - 2s/step
Epoch 148/150
33/33 - 63s - loss: 0.0447 - accuracy: 0.9844 - val_loss: 1.6082 - val_accuracy:
0.7072 - 63s/epoch - 2s/step
Epoch 149/150
33/33 - 63s - loss: 0.1083 - accuracy: 0.9659 - val_loss: 2.3715 - val_accuracy:
0.6578 - 63s/epoch - 2s/step
Epoch 150/150
33/33 - 63s - loss: 0.1440 - accuracy: 0.9503 - val_loss: 1.5047 - val_accuracy:
0.7262 - 63s/epoch - 2s/step
```

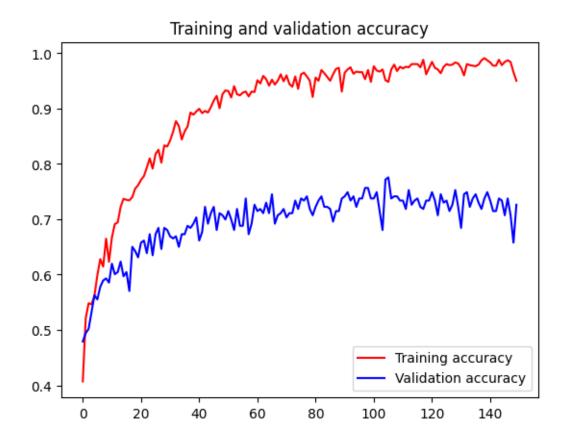


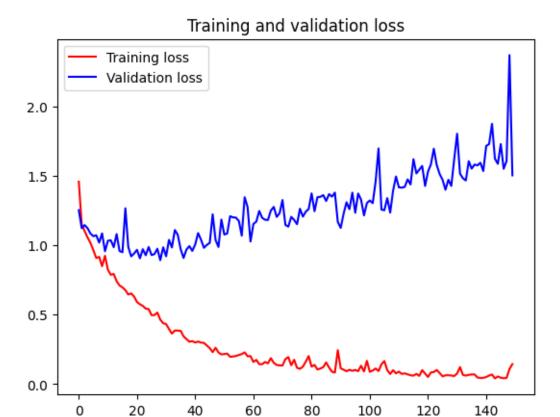
[30]: pd.DataFrame(history.history).plot(figsize = (10, 7), xlabel = "epochs");



```
[31]: # #plot the learning rate versus the loss
     # lrs = 1e-4 * (10 ** (tf.range(300)/20))
     # plt.figure(figsize = (10,7))
     # plt.semilogx(lrs, history.history["loss"])
     # plt.xlabel("learning rate")
     # plt.ylabel("loss")
     # plt.title("learning rate vs loss")
[32]: %matplotlib inline
     import matplotlib.image as mpimg
     import matplotlib.pyplot as plt
[33]: # evaluating accuracy and loss for the model
[34]: #-----
     # Retrieve a list of list results on training and test data
     # sets for each training epoch
     #-----
           = history.history[ 'accuracy']
     val_acc = history.history[ 'val_accuracy' ]
     loss = history.history[ 'loss']
     val_loss = history.history['val_loss' ]
     epochs = range(len(acc)) # Get number of epochs
     # Plot training and validation accuracy per epoch
     plt.plot(epochs, acc, 'r', label='Training accuracy')
     plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
     plt.title ('Training and validation accuracy')
     plt.legend(loc=0)
     plt.figure()
     #-----
     # Plot training and validation loss per epoch
     #-----
     plt.plot ( epochs, val loss , 'b', label='Validation loss')
     plt.title ('Training and validation loss' )
     plt.legend(loc=0)
     plt.figure()
```

[34]: <Figure size 640x480 with 0 Axes>





<Figure size 640x480 with 0 Axes>

3 menyimpan model TAPI BELUM YANG VERSI QUANTIZED

```
[35]: from keras.models import load_model
    model.save('/home/jupyter/content/kerasFormat_model5/model_vggFace_13Juni.h5')

[36]: from keras.models import load_model
    model.save('/home/jupyter/content/saved Model5')

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op,
```

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 16). These functions will not be directly callable after loading.

 ${\tt INFO: tensorflow: Assets \ written \ to: /home/jupyter/content/saved \ Model5/assets}$

INFO:tensorflow:Assets written to: /home/jupyter/content/saved Model5/assets

4 Convert to Tensorflow LITE

```
[]: try:
         %tensorflow_version 2.x
     except:
         pass
[]: import pathlib
     print('\u2022 Using TensorFlow Version:', tf.__version__)
[]:
[]: | #!mkdir saved_TFLITE_model
    generate SavedModel
[]: export_dir = "home/jupyter/content/saved_TFLITE_model"
     tf.saved_model.save(model, export_dir)
[]: model = tf.saved_model.load(export_dir)
    Convert the SavedModel to JSON
[]: #!pip install --user tensorflowjs
[]: # import tensorflowjs
     # import json
[]: # # Load the SavedModel
     # model = tf.saved_model.load('home/jupyter/content/saved Model')
[]: import tensorflow as tf
     import json
     # Load the SavedModel
     model = tf.saved_model.load(export_dir)
     # Convert tensor shapes to lists
     def convert shape(shape):
         return [dim for dim in shape.as_list()]
     # Create a dictionary to store the JSON model
     json_model = {}
     # Get information about inputs
     input_signatures = model.signatures['serving_default'].
      ⇒structured_input_signature[1]
     json_model['inputs'] = []
```

```
for tensor_name, tensor_info in input_signatures.items():
    input_info = {
        'name': tensor_name,
        'dtype': str(tensor_info.dtype),
        'shape': convert_shape(tensor_info.shape),
    json_model['inputs'].append(input_info)
# Get information about outputs
output_signatures = model.signatures['serving_default'].structured_outputs
json model['outputs'] = []
for tensor_name, tensor_info in output_signatures.items():
    output info = {
        'name': tensor_name,
        'dtype': str(tensor_info.dtype),
        'shape': convert_shape(tensor_info.shape),
    }
    json_model['outputs'].append(output_info)
# Save the JSON model to a file
with open('model.json', 'w') as f:
    json.dump(json_model, f)
```

Convert the SavedModel to TFLite

```
[]: # tflite_model_file = pathlib.Path('home/jupyter/content/')
# tflite_model_file.write_bytes(tflite_model)
```

5 labels

```
[]: class_names = ['combination', 'dry', 'nonface', 'normal', 'oily', 'sensitive']

create a txt file to save the 5 labels
```

```
[]: with open('faces_labels.txt', 'w') as f:
        f.write('\n'.join(class_names))
[]: import shutil
    shutil.copy("/content/faces_labels.txt","/content/drive/MyDrive/saved model")
[]: from google.colab import drive
    drive.mount('/content/drive')
[]: import os
    file_path = "/home/jupyter/content/kerasFormat_model/model_vggFace_10Juni.h5" __
      →# Replace with the correct file path
    file_size = os.path.getsize(file_path)
    print(file_size)
[]: | wget -0 /home/jupyter/content/kerasFormat_model/model_vggFace_10Juni.h5 https:/
      △/58948e97954de7e0-dot-asia-southeast2.notebooks.googleusercontent.com/files/
      ⇔content/kerasFormat_model/model_vggFace_10Juni.h5?
      ssrf=2%7C1f091a48%7C9c97e6fd85cef3a278dc9a8e0d4188fa%7C1686143743
[]:
```