Programmierung für KI Projektausarbeitung - Haar Cascades

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Einleitung

Im Rahmen der Veranstaltung - *Programmierung für KI*, haben wir das Thema **Haar Cascades**, als gemeinsam zu bearbeitendes Projekt erhalten. Haar Cascades beschreibt ein Verfahren bzw. einen Algorithmus, welches bestimmte Merkmale oder Muster in Bildern oder Videos erkennen kann.

Ziel dieses Projektes ist es nun ein solches Verfahren zu implementieren und eine passende graphische Benutzeroberfläche in Python bereitzustellen. Nach einer kurzen theoretischen Einführung in das Thema, teilt sich die Arbeit in zwei Teile auf. Der erste Teil beschäftigt sich mit der Erstellung einer graphischen Benutzeroberfläche mit Hilfe von **Tkinter** und der zweite Teil mit **steamlit**.

Schrittweise werden wir auf die einzelnen Python Codes eingehen und diese näher erläutern. Des Weiteren werden wir kennzeichnen, welche Codeblock genau, welchem Projektteilnehmer zuzuordnen sind.

1 Einführung - Haar Cascades

Wie bereits oben erwähnt sind Haar Cascades ¹ ein Verfahren zur Erkennung von Objekten in Bildern und Videos. Genauer handelt es sich hierbei um einen Klassifikationsalgorithmus, welches Beispielbilder trainiert, die entweder das zu erkennende Muster oder Objekt enthalten oder nicht. Anschließend kann der Haar Cascade dann auf neue Bilder oder Videos angewendet werden und versucht, das gesuchte Muster oder Objekt darin wieder zu erkennen.

1.1 Funktionsweise: Haar-like Features

Aus den Bildern als Eingangsdaten werden so genannte Haar-like Features als Merkmale extrahiert. Für diese Merkmale werden rechteckige Bereiche im Bild zusammengefasst.

¹Rapid object detection using a boosted cascade of simple features und Viola, Jones: Robust Real-time Object Detection, IJCV 2001, zurück zu führen auf *Paul Viola* und *Michael Jones*

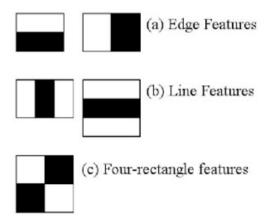


Abbildung 1: Haar-like Features

1.2 Umsetzung in OpenCV

Für unseren Anwendungsfall ist es mit Hilfe von OpenCV nicht notwendig einen eigenen Haar Cascade Algorithmus anzulernen. Wir laden die XML-Datei mit den zugehörigen *Haar-like-Features*, die wir verwenden möchten. Diese XML-Dateien enthalten Informationen darüber, wie die Features bzw. Merkmale aussehen und wie sie auf Bilder angewendet werden sollen.

2 Implementation in Tkinter

Um den Code möglichst lesbar zu gestalten, teilen wir diesen im Folgenden in mehreren kleinen Blöcke auf.

2.1 Packages

```
import tkinter as tk
from tkinter import ttk
from tkinter import filedialog
import numpy as np
import cv2
from PIL import Image, ImageTk
import random
import os

from pki_a22_app.utils.file_loader import get_classifiers

path_haarcascade = "resources/haarcascades/haarcascade_"

classifier_list = get_classifiers()
```

2.2 Initialisierung und Geometrie

```
root = tk.Tk()
2 root.title('Projekt Gruppe a2-2 - Thema: Bilderkennung Haar-
      Cascades')
5 screen_width = root.winfo_screenwidth()
6 screen_height = root.winfo_screenheight()
9 if screen_width/screen_height < 1.8:</pre>
window_width = int(screen_width * 0.8)
11 else:
   window_width = int(screen_height * (screen_width/2/
12
      screen_height)*0.8)
   window_height = int(screen_height * 0.8)
13
img_max_width = int(window_width/2-75)
img_max_height = int(window_height-300)
17
center_x = int(screen_width/2 - window_width / 2)
20 center_y = int(screen_height/2 - window_height / 2)
22
23
24 root.geometry(f'{window_width}x{window_height}+{center_x}+{
      center_y}')
25 root.resizable(False, False)
```

2.3 Bild öffnen und Klassifizierer anwenden

```
def file_open():
2
    global input_image
    global output_image
    filename = filedialog.askopenfilename(filetypes=(("jpg files",
      "*.jpg"),("png files", "*.png")))
    input_image = Image.open(filename)
    width, height = input_image.size
6
    aspect_ratio = width / height
    if aspect_ratio > 1:
     new_width = img_max_width
      new_height = int(new_width / aspect_ratio)
10
    new_height = img_max_height
12
    new_width = int(new_height * aspect_ratio)
13
14
    input_image = input_image.resize((new_width,new_height), Image.
      ANTIALIAS)
15
    input_image_tk = ImageTk.PhotoImage(input_image)
    input_img_label.config(image=input_image_tk)
16
    input_img_label.image = input_image_tk
17
18
19
    output_image = Image.open(filename)
    output_image = output_image.resize((new_width,new_height),
20
     Image.ANTIALIAS)
21
    output_image_tk = ImageTk.PhotoImage(output_image)
    output_img_label.config(image=output_image_tk)
22
    output_img_label.image = output_image_tk
def img_change(classifier):
    global input_image
    global output_image
3
    cascade = cv2.CascadeClassifier(path_haarcascade + classifier +
       ".xml")
    output_image_cv = np.array(output_image.convert('RGB'))
    output_image_cv_gray = cv2.cvtColor(output_image_cv, cv2.
      COLOR_BGR2GRAY)
    cascade_results = cascade.detectMultiScale(output_image_cv_gray
      , scaleFactor=s1.get_val(), minNeighbors = s2.get_val(),
      minSize=(s3.get_val(), s3.get_val()))
    iterations = 0
    if len(cascade_results) > 0:
      color = (random.randint(0,255),random.randint(0,255),random.
11
      randint(0,255))
12
      for (x,y,w,h) in cascade_results:
        cv2.rectangle(output_image_cv,(x,y),(x+w,y+h),(color),2)
        roi_gray = output_image_cv_gray[y:y+h, x:x+w]
        roi_color = output_image_cv[y:y+h, x:x+w]
      output_image = Image.fromarray(output_image_cv)
16
      output_img_label.image.paste(output_image)
17
    else:
18
      flag = False
19
      for i1 in np.arange (1.5, 0.9, -0.1):
20
        for i2 in range (6, 2, -1):
21
          for i3 in range (50, 9, -10):
22
        s1.s.<u>set</u>(i1)
23
```

```
s2.s.set(i2)
24
            s3.s.set(i3)
            iterations = iterations + 1
26
            cascade_results = cascade.detectMultiScale(
      output_image_cv_gray , scaleFactor=s1.get_val() , minNeighbors
      = s2.get_val(), minSize=(s3.get_val(), s3.get_val()))
             if len(cascade_results) > 0:
              color = (random.randint(0,255),random.randint(0,255),
29
      random.randint(0,255))
              for (x,y,w,h) in cascade_results:
30
                 cv2.rectangle(output_image_cv,(x,y),(x+w,y+h),(
31
      color),2)
                roi_gray = output_image_cv_gray[y:y+h, x:x+w]
                roi_color = output_image_cv[y:y+h, x:x+w]
                 output_image = Image.fromarray(output_image_cv)
34
                 output_img_label.image.paste(output_image)
36
                flag = True
                break
37
            if flag: break
          if flag: break
39
        if len(cascade_results) == 0:
         print(f"Kein Ergebnis nach {iterations} Iterationen")
41
def output_image_restart():
  global input_image
    global output_image
3
    output_image = input_image
    output_img_label.image.paste(output_image)
```

2.4 Slider und Button

```
1 class slider:
   def __init__(self,name,x_pos=0,y_pos=0,scale_from=0,scale_to
      =100, typ=int):
      self.name = name
      self.x_pos = x_pos
4
      self.y_pos = y_pos
      self.scale_from = scale_from
6
      self.scale_to = scale_to
      if typ==int:
        self.val = tk.IntVar()
9
      else:
10
       self.val = tk.DoubleVar()
11
12
      self.lbl = ttk.Label(root, text=name)
13
      self.lbl.place(x=self.x_pos,y=self.y_pos)
14
      self.s = ttk.Scale(root, from_=self.scale_from, to=self.
16
      scale_to, orient="horizontal",command=self.slider_change,
      variable=self.val)
      self.s.place(x=self.x_pos+100,y=self.y_pos)
17
      self.v = ttk.Label(root,text=f"{self.get_val():.1f}")
19
      self.v.place(x=self.x_pos+210,y=self.y_pos)
    def get_val(self):
21
  return self.val.get()
```

```
def slider_change(self, event):
self.v.configure(text=f"{self.get_val():.1f}")
def blur_rectangle(classifier):
    global input_image
    global output_image
    cascade = cv2.CascadeClassifier(path_haarcascade + classifier +
       ".xml")
    output_image_cv = np.array(output_image.convert('RGB'))
    output_image_cv_gray = cv2.cvtColor(output_image_cv, cv2.
      COLOR_BGR2GRAY)
    cascade_results = cascade.detectMultiScale(output_image_cv_gray
      , scaleFactor=s1.get_val(), minNeighbors = s2.get_val(),
      minSize=(s3.get_val(), s3.get_val()))
      if len(cascade_results) > 0:
        for (x,y,w,h) in cascade_results:
10
          face = output_image_cv[y:y+h, x:x+w]
          face = cv2.GaussianBlur(face, (23, 23), 30)
output_image_cv[y:y+h, x:x+w] = face
12
13
        output_image = Image.fromarray(output_image_cv)
14
        output_img_label.image.paste(output_image)
1.5
16
17 def save_jpg():
    file_path = filedialog.asksaveasfilename(initialfile="
      output_image", filetypes=(("jpg files", "*.jpg"),("png files"
        "*.png")), defaultextension=".jpeg")
    if file_path:
19
20
      output_image.save(file_path)
21
22 tk.Button(root, text="Bild aus Datei oeffnen", command=file_open)
      .place(x=window_width/2-window_width/4,y=150)
23 tk.Button(root, text="Classifier anwenden", command=lambda:
      img_change(dropdown.get())).place(x=window_width/2+
      window_width/4, y=150)
24 tk.Button(root, text="==>", command=output_image_restart).place(x
      =window_width/2-12.5,y=window_height/2)
tk.Button(root, text="Weichzeichnen",command=lambda:
      blur_rectangle(dropdown.get())).place(x=window_width/2+
      window_width/4+122, y=150)
26 tk.Button(root, text="Bild speichern",command=save_jpg).place(x=
      window_width/2+window_width/4+220, y=150)
27
28 dropdown = tk.StringVar(root)
29 dropdown.set(classifier_list[2])
30 dropdown_label = tk.OptionMenu(root, dropdown, *classifier_list)
dropdown_label.place(x=window_width/2-75,y=20)
xpos_slider_window = window_width/2 -75
35 ypos_slider_window = 60
36 s1 = slider("ScaleFactor", xpos_slider_window, ypos_slider_window
       ,1.01,1.5,float)
37 s1.s.set(1.1)
38 s2 = slider("MinNeighbors", xpos_slider_window, ypos_slider_window
      +30,3,6)
39 s2.s.set(4)
```

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3 Implementierung mit Streamlit

Eine weitere Implementierung des Haar Cascade Klassifizierers wurde mit dem Dashboard Framework **Streamlit** als **Web-Anwendung** umgesetzt. Das Dashboard enthält zum einen einen Bereich mit unterschiedlichen Parametern für den Haar Cascade Klassifizierer, und zum anderen einen Bereich, in dem verschieden Bild-Quellen ausgewählt werden können. Dazu gehört neben eigenen oder vorinstallierten Bildern auch die Möglichkeit, ein Video oder das eigene Webcam-Bild zu verwenden.

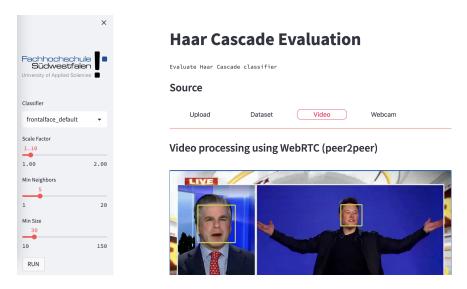


Abbildung 2: Streamlit Dashboard mit Haar Cascade Klassifizierer

Das Dashboard kann auch als https://jk-fhswf-pki-a22-app-app-codcuk.streamlit.app/Demo-Anwendung auf der Streamlit-Cloud ausprobiert werden.

3.1 Eingabequellen

Die Anwendung unterstützt diese bereits angesprochenen Eingabemöglichkeiten:

Upload

Hier kann ein einzelnes Bild hochgeladen und verwendet werden

Dataset

Unter diesem Punkt können mehrere vorkonfigurierte Bilder auf einmal verwendet werden

Video

Bei dieser Option wird ein vorkonfiguriertes Video abgespielt, auf dem der ausgewählte Klassifizierer angewandt wird

Webcam

Hier besteht die Möglichkeit, den Klassifizierer auf die eigene Webcam-Bilder anzuwenden

3.2 Optionen

Die verwendeten Optionen zur Konfiguration des Haar Cascade Klassifizierers Scale Factor, Min Neighbors und Min Size wurden bereits in den vorherigen Abschnitten im Rahmen der Tkinter Anwendung beschrieben.

3.3 Code

Der Programmcode für die Implementierung des Streamlit-Dashboards ist auf die folgenden Quelldateien verteilt:

```
1 # Hauptanwendung und Einstiegspunkt
2 app.py
3 # Funktion zur Ausfuehrung des eigentlichen Klassifizierers
4 pki_a22_app/haarcascades/haarcascades.py
   Interface fuer die unterschiedlichen Eingabequellen
6 pki_a22_app/dashboard/sources.py
  # Eingabequelle mit einem einzelnen hochladbarem Bild
8 pki_a22_app/dashboard/source_upload.py
  # Eingabequelle mit Auswahl eines Datensatzes mit mehreren
      Bildern
pki_a22_app/dashboard/source_dataset.py
11 # Eingabequelle zum Abspielen eines voreingestellten Videos
pki_a22_app/dashboard/source_video.py
# Eingabequelle bei der die eigene Webcam verwendet wird
pki_a22_app/dashboard/source_webcam.py
15 # Utility Modul, das vor allem Funktionen fuer Datei-Operationen
      enthaelt
pki_a22_app/utils/file_loader.py
```

4 Organisation

Besonders zu Beginn war es sehr hilfreich, eine Kollaborations-Plattform für das Projekt einzurichten, über die die unterschiedlichsten Informationen organisatorischer Natur ausgetauscht werden konnten. Wir haben uns für den Einsatz eines Miro-Boardes entschieden. das wir unter anderem nutzten für:

- Terminfindung
- Pinnwand
- Brainstorming
- Zeitplanung
- Wireframes

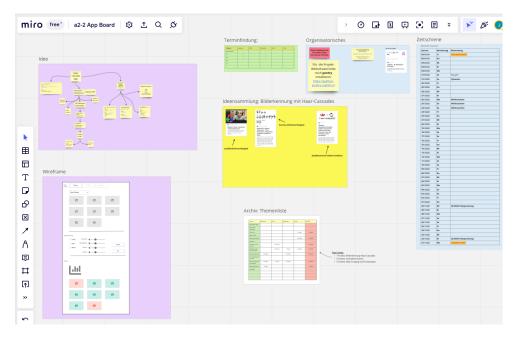


Abbildung 3: Miro-Board für organisatorische Aufgaben

5 Continuos Integration

Um gemeinsam am Programmcode arbeiten zu können, haben wir uns ein Gitbasiertes Codeverwaltungs-Tool entschieden. Hierfür haben wir auf GitHub ein eigenes https://github.com/jk-fhswf/pki-a22-appGit-Repository eingerichtet.

Zur Qualitätssicherung haben wir uns dafür entschieden, für den main-Branch Pull-Requests zu verwenden und die Erfolgreiche Ausführung eines GitHub-Actions- Workflows für Unit-Tests vorauszusetzen.

6 Anhang Streamlit Code

```
"""Start a streamlit dashboard showing some haarcascade
operations with Opency"""

# Core Pkgs

import extra_streamlit_components as stx
import streamlit as st
from pki_a22_app.dashboard.source_dataset import DatasetSource
from pki_a22_app.dashboard.source_upload import UploadSource
from pki_a22_app.dashboard.source_video import VideoSource
from pki_a22_app.dashboard.source_webcam import WebcamSource
```

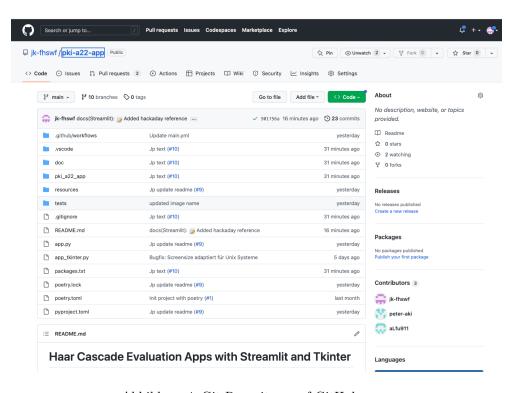


Abbildung 4: Git-Repository auf GitHub

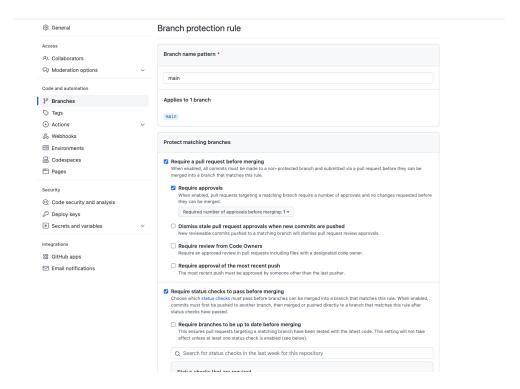


Abbildung 5: Konfiguration von QA-Optionen

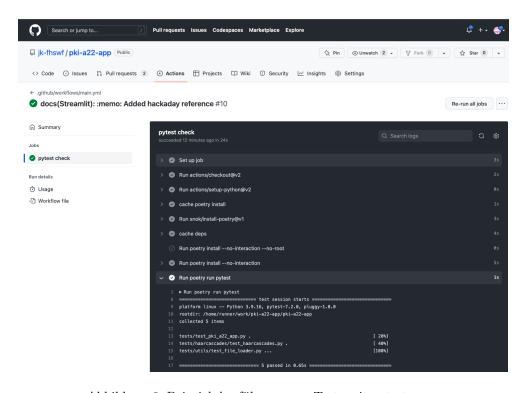


Abbildung 6: Beispiel Ausführung von Tests mit pytest

```
12 from pki_a22_app.utils.file_loader import (get_classifiers)
13
14
def main():
16
      Main function that initiates the Streamlit dashboard
17
18
19
      st.title("Haar Cascade Evaluation")
20
      st.text("Evaluate Haar Cascade classifier")
21
22
      # Configure the sources section
23
      st.subheader("Source")
24
      chosen_id = stx.tab_bar(data=[
          {\tt stx.TabBarItemData(id=1,\ title="Upload",\ description=""),}
26
          stx.TabBarItemData(id=2, title="Dataset", description="")
          stx.TabBarItemData(id=3, title="Video", description=""),
28
          stx.TabBarItemData(id=4, title="Webcam", description=""),
      ], default=1)
30
      st.sidebar.image("resources/images/logo-fh-swf-300x93.png")
32
      st.sidebar.markdown('#')
33
34
      # Configure the algorithm controls
35
      classifiers: list = get_classifiers()
      classifier_id = st.sidebar.selectbox("Classifier",
37
      classifiers)
      scale_factor = st.sidebar.slider('Scale Factor', 1.0, 2.0,
38
      1.1, 0.1)
      min_neighbors = st.sidebar.slider('Min Neighbors', 1, 20, 5,
      1)
      min_size = st.sidebar.slider('Min Size', 10, 150, 30, 5)
40
41
      show_results = st.sidebar.button("RUN")
42
43
      # Define different sources that render the input and output
44
      accordingly
      sources = {
45
          "1": UploadSource(),
46
          "2": DatasetSource(),
47
          "3": VideoSource(),
48
          "4": WebcamSource()
49
50
      # Execute selected source
52
      sources[chosen_id].load_source(
53
          classifier_id, scale_factor, min_neighbors, min_size,
54
      show_results)
56
if __name__ == '__main__':
main()
```

Listing 1: app.py

```
1 """This module contains haar cascade core functionality"""
```

```
3 import cv2
  4 from cv2 import Mat
  5 import numpy.typing as npt
  8 def detect_objects(np_input_image: npt.ArrayLike, classifier_id:
                     str, scale_factor: float = 1.1, min_neighbors: int = 4,
                     min_size: int = 10) -> Mat:
10
                     Detects objects in an image using Haar Cascades classifiers.
                     It can usa a preset of trained classifiers in resources/
                     haarcascades/*.
                     The found objects get highlighted by a rectangle around it.
12
                     The parameter description was taken from:
13
                     https://hackaday.io/project/12384-autofan-automated-control-
14
                     of-air-flow/log/41956-face-detection-using-a-haar-cascade-
                     classifier
16
                     Parameters
18
                      _____
                     np_input_image : npt.ArrayLike
19
20
                                  The input image
                     classifier_id : str
21
                                 the classifier name. E.g. when you set it to "
22
                     frontalface_default", the configuration file
                                  "haarcascade_frontalface_default.xml" will be loaded
23
                     scale_factor : float, optional, by default 1.1
24
                                  determines the factor by which the detection window of
25
                     the classifier is scaled down per detection pass. A factor of
                                  corresponds to an increase of 10%. Hence, increasing the
                     scale factor increases performance, as the number of % \left( 1\right) =\left( 1\right) \left( 1\right) 
                     detection
                                   passes is reduced. However, as a consequence the
                     reliability by which a face is detected is reduced. See:
                     min_neighbors : int, optional, by default 4
                                   determines the minimum number of neighboring facial
29
                      features that need to be present to indicate the detection of
                        a face by the
                                   classifier. Decreasing the factor increases the amount of
30
                        false positive detections. Increasing the factor might lead
                     to missing
                                   faces in the image. The argument seems to have no
31
                     influence on the performance of the algorithm. See:
                     min_size: int
                                  determines the minimum size of the detection window in
                     pixels. Increasing the minimum detection window increases
                     performance.
                                 However, smaller faces are going to be missed then. In
                     the scope of this project however a relatively big detection
                     window can
                               be used, as the user is sitting directly in front of the
35
                      camera.
36
                     Returns
```

```
38
39
      cv2.Mat
        The image in Mat representation
40
41
      A list of found objects in the image
42
43
44
      classifier = cv2.CascadeClassifier(
45
          f"resources/haarcascades/haarcascade_{classifier_id}.xml"
47
      img = cv2.cvtColor(np_input_image, 1)
      gray = cv2.cvtColor(np_input_image, cv2.COLOR_BGR2GRAY)
49
      objects = classifier.detectMultiScale(
50
          gray, scaleFactor=scale_factor, minNeighbors=
51
      min_neighbors, minSize=(min_size, min_size))
      # pylint: disable=invalid-name
52
      for (x, y, w, h) in objects:
53
          cv2.rectangle(img, (x, y), (x + w, y + h), (72, 209, 204)
54
      return img, objects
```

Listing 2: haarcascades/haarcascades.py

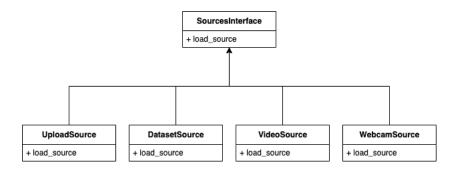


Abbildung 7: UML Diagramm für unterschiedliche Source-Typen

```
1 """
2 Contains an interface definition to use different source-types
_{\rm 3} in the Streamlit dashboard
5 import abc
8 class SourcesInterface(metaclass=abc.ABCMeta):
9
10
      Interface for handling different input source types
11
      Methods
13
      load_source(classifier_id: str, scale_factor: int,
14
      min_neighbors: int, show_results: bool = False)
         Load the sources and outputs with respect to the given
      parameters
16
17
      @classmethod
18
      def __subclasshook__(cls, subclass):
19
20
          Ensures that needed methods are implemented in your
21
      derivatives
22
          Parameters
23
24
           subclass : Any
25
             subclass that was created
27
28
          Returns
29
           Bool
30
              True if it is a valid subclass
31
32
           return (hasattr(subclass, 'load_source') and
                   callable(subclass.load_source) or
34
                   NotImplemented)
35
36
37
      @abc.abstractmethod
```

```
def load_source(self, classifier_id: str, scale_factor: int,
      min_neighbors: int, min_size: int, show_results: bool = False
      ):
40
          This method handles the display of different Streamlit-UI
41
       elements. Implement these in your
          concrete derivatives
43
          The parameter description was taken from:
          https://hackaday.io/project/12384-autofan-automated-
45
      control-of-air-flow/log/41956-face-detection-using-a-haar-
      cascade-classifier
46
          Parameters
47
48
          classifier_id : str
49
50
             the classifier name. E.g. when you set it to "
      frontalface_default", the configuration file
              "haarcascade_frontalface_default.xml" will be loaded
51
          scale\_factor : float, optional, by default 1.1
52
              determines the factor by which the detection window
      of the classifier is scaled down per detection pass. A factor
              corresponds to an increase of 10%. Hence, increasing
54
      the scale factor increases performance, as the number of
      detection
              passes is reduced. However, as a consequence the
      reliability by which a face is detected is reduced. See:
          min_neighbors : int, optional, by default 4
56
              determines the minimum number of neighboring facial
      features that need to be present to indicate the detection of
       a face by the
              classifier. Decreasing the factor increases the
      amount of false positive detections. Increasing the factor
      might lead to missing
              faces in the image. The argument seems to have no
      influence on the performance of the algorithm. See:
          min_size: int
              determines the minimum size of the detection window
61
      in pixels. Increasing the minimum detection window increases
      performance.
              However, smaller faces are going to be missed then.
62
      In the scope of this project however a relatively big
      detection window can
              be used, as the user is sitting directly in front of
63
      the camera.
          show_results : bool, optional
64
              If true we want to display the original images AND
      the results, by default False
67
          Raises
68
69
          NotImplementedError
             Error that gets thrown if you do not implement this
70
      function
```

72 raise NotImplementedError

Listing 3: dashboard/sources.py

```
1 """Implementation of a source-type to upload an image and apply
      the Haar Cascade classifier on it"""
2 # Core Pkgs
4 import numpy as np
5 import streamlit as st
6 from PIL import Image
8 from pki_a22_app.dashboard.sources import SourcesInterface
9 from pki_a22_app.haarcascades.haarcascades import detect_objects
11
class UploadSource(SourcesInterface):
      def load_source(self, classifier_id: str, scale_factor: int,
13
      min_neighbors: int, min_size: int, show_results: bool = False
14
          This method creates a form to upload an image that can be
       processed by the haar cascade classifier afterwards.
16
          The parameter description was taken from:
17
          https://hackaday.io/project/12384-autofan-automated-
18
      control-of-air-flow/log/41956-face-detection-using-a-haar-
      cascade-classifier
19
          Parameters
20
21
          classifier_id : str
              the classifier name. E.g. when you set it to "
23
      frontalface_default", the configuration file
              "haarcascade_frontalface_default.xml" will be loaded
24
          scale_factor : float, optional, by default 1.1
25
              determines the factor by which the detection window
      of the classifier is scaled down per detection pass. A factor
              corresponds to an increase of 10%. Hence, increasing
27
      the scale factor increases performance, as the number of
      detection
              passes is reduced. However, as a consequence the
      reliability by which a face is detected is reduced. See:
          \min\_neighbors : int, optional, by default 4
29
              determines the minimum number of neighboring facial
30
      features that need to be present to indicate the detection of
       a face by the
              classifier. Decreasing the factor increases the
      amount of false positive detections. Increasing the factor
      might lead to missing
              faces in the image. The argument seems to have no
      influence on the performance of the algorithm. See:
33
          min_size: int
              determines the minimum size of the detection window
34
      in pixels. Increasing the minimum detection window increases
      performance.
```

```
However, smaller faces are going to be missed then.
35
      In the scope of this project however a relatively big
      detection window can
              be used, as the user is sitting directly in front of
      the camera.
          show_results : bool, optional
37
              If true we want to display the original images AND
      the results, by default False
          st.subheader("Upload an image")
40
41
          # Show the upload form
42
          image_file = st.file_uploader(
43
               "Upload Image", type=["jpg", "png", "jpeg"])
44
          our_image = None
45
46
          # If a image was uploaded, display it
47
          if image_file is not None:
48
              our_image = Image.open(image_file)
49
              our_np_image = np.array(our_image.convert("RGB"))
50
               st.text("Original Image")
51
              st.image(our_image)
52
              image_location = st.empty()
53
54
          # Show processed image
55
          if show_results and classifier_id:
56
               # Actual object detection
57
               result_img, result_faces = detect_objects(
58
                   our_np_image, classifier_id, scale_factor,
59
      min_neighbors, min_size)
               image_location.image(result_img)
60
               st.text("Resulting Image")
61
               st.success(f"Found {len(result_faces)} Objects")
62
```

Listing 4: dashboard/source upload.py

```
1 """Implementation of a source-type that show multiple images at
       once"""
3 # Core Pkgs
4 from itertools import cycle
6 import numpy as np
7 import streamlit as st
8 from PIL import Image
{\tt 10} \  \  \, \textbf{from} \  \  \, \textbf{pki_a22\_app.dashboard.sources} \  \  \, \textbf{import} \  \  \, \textbf{SourcesInterface}
11 from pki_a22_app.haarcascades.haarcascades import detect_objects
from pki_a22_app.utils.file_loader import get_datasets,
       get_images_of_dataset
13
14
class DatasetSource(SourcesInterface):
16
       Implementation for handling multiple images, also called
17
       dataset here
18
   Methods
```

```
20
      load_source(classifier_id: str, scale_factor: int,
      min_neighbors: int, show_results: bool = False)
         Load multiple images and display results
22
24
25
      def load_source(self, classifier_id: str, scale_factor: int,
      min_neighbors: int, min_size: int, show_results: bool = False
26
          This method shows a selectbox to select an existing
      dataset. A dataset contains multiple images. The datasets are
       located
          at resource/datasets/. The Haar Cascade classifier will
      then be applied to all images of the dataset.
29
30
          The parameter description was taken from:
          https://hackaday.io/project/12384-autofan-automated-
31
      control-of-air-flow/log/41956-face-detection-using-a-haar-
      cascade-classifier
          Parameters
33
34
          classifier_id : str
35
              the classifier name. E.g. when you set it to "
36
      frontalface_default", the configuration file
              "haarcascade_frontalface_default.xml" will be loaded
37
          scale_factor : float, optional, by default 1.1
38
              determines the factor by which the detection window
39
      of the classifier is scaled down per detection pass. A factor
              corresponds to an increase of 10%. Hence, increasing
40
      the scale factor increases performance, as the number of
      detection
              passes is reduced. However, as a consequence the
41
      reliability by which a face is detected is reduced. See:
          min_neighbors : int, optional, by default 4
42
              determines the minimum number of neighboring facial
      features that need to be present to indicate the detection of
       a face by the
              classifier. Decreasing the factor increases the
      amount of false positive detections. Increasing the factor
      might lead to missing
             faces in the image. The argument seems to have no
45
      influence on the performance of the algorithm. See:
          min_size: int
46
              determines the minimum size of the detection window
47
      in pixels. Increasing the minimum detection window increases
      performance.
              However, smaller faces are going to be missed then.
      In the scope of this project however a relatively big
      detection window can
49
             be used, as the user is sitting directly in front of
      the camera.
          show_results : bool, optional
50
             If true we want to display the original images AND
51
      the results, by default False
```

```
52
53
          st.subheader("Load an existing dataset")
54
          # Setup dataset selector
          datasets = get_datasets()
56
          datasets.insert(0, "None")
57
          dataset_selection = st.selectbox("Dataset", datasets)
58
59
          img_width = 300
60
          img_columns = 2
61
62
          # When a dataset was selected start presenting it
63
          if dataset_selection and dataset_selection != "None":
64
               images_in_ds = get_images_of_dataset(
65
      dataset_selection)
66
67
               # Show images in max n columns
               cols = cycle(st.columns(img_columns))
68
69
               for idx, filtered_image in enumerate(images_in_ds):
                   next(cols).image(str(filtered_image), width=
70
      img_width)
71
               # We were requested to show the results
72
73
               if show_results and dataset_selection:
                  images_in_ds = get_images_of_dataset(
74
      dataset_selection)
                   st.write("### Results:")
                   cols = cycle(st.columns(img_columns))
76
77
                   for idx, filtered_image in enumerate(images_in_ds
78
      ):
                       ds_image = Image.open(filtered_image)
79
                       ds_np_image = np.array(ds_image.convert("RGB"
80
      ))
                       # Actual object detection
81
                       ds_res_image, ds_result_faces =
      detect_objects(ds_np_image, classifier_id, scale_factor,
        min_neighbors, min_size)
                       col = next(cols)
84
                       col.image(ds_res_image, width=img_width,
      caption=f"{len(ds_result_faces)} objects")
```

Listing 5: dashboard/source dataset.py

```
12
13
14 class VideoSource(SourcesInterface):
      def load_source(self, classifier_id: str, scale_factor: int,
      min_neighbors: int, min_size: int, show_results: bool = False
          This function shows a video player where the haar cascade
       detection is applied on single frames.
          The player uses WebRTC to stream the video.
18
19
20
          The parameter description was taken from:
          https://hackaday.io/project/12384-autofan-automated-
21
      control-of-air-flow/log/41956-face-detection-using-a-haar-
      cascade-classifier
22
23
          Parameters
24
25
          classifier id : str
26
              the classifier name. E.g. when you set it to "
      frontalface_default", the configuration file
              "haarcascade_frontalface_default.xml" will be loaded
28
29
          scale_factor : float, optional, by default 1.1
              determines the factor by which the detection window
30
      of the classifier is scaled down per detection pass. A factor
       of 1.1
              corresponds to an increase of 10%. Hence, increasing
31
      the scale factor increases performance, as the number of
      detection
              passes is reduced. However, as a consequence the
      reliability by which a face is detected is reduced. See:
          \min\_neighbors : int, optional, by default 4
33
              determines the minimum number of neighboring facial
34
      features that need to be present to indicate the detection of
       a face by the
              classifier. Decreasing the factor increases the
35
      amount of false positive detections. Increasing the factor
      might lead to missing
              faces in the image. The argument seems to have no
      influence on the performance of the algorithm. See:
          min_size: int
              determines the minimum size of the detection window
      in pixels. Increasing the minimum detection window increases
      performance.
39
              However, smaller faces are going to be missed then.
      In the scope of this project however a relatively big
      detection window can
              be used, as the user is sitting directly in front of
40
      the camera.
          show_results : bool, optional
41
              If true we want to display the original images AND
42
      the results, by default False
43
          st.subheader("Video processing using WebRTC (peer2peer)")
45
          # Callback function that takes an image frame that can
```

```
# be processed
47
           def callback(frame):
               # convert frame to ndarray
49
               img = frame.to_ndarray(format="bgr24")
5.1
               # actual object detection
52
               result_img, result_faces = detect_objects(
53
                   img, classifier_id, scale_factor, min_neighbors,
54
      min_size)
               return av.VideoFrame.from_ndarray(result_img, format=
56
      "bgr24")
57
           rtc_configuration = {  # Add this config
58
               "iceServers": [{"urls": ["stun:stun.1.google.com
59
       :19302"]}]
60
          }
61
62
           def create_player():
               return MediaPlayer("resources/videos/musk.mp4")
63
64
           # This will show the video-UI elements in Streamlit
65
           # Each frame gets sent to the callback method
66
           webrtc_streamer(key="videoexample",
67
                           video_frame_callback=callback,
68
69
                           rtc_configuration=rtc_configuration,
                           mode=WebRtcMode.RECVONLY,
70
                           player_factory=create_player,
71
72
                           media_stream_constraints={
                                "video": True
73
74
75
```

Listing 6: dashboard/source video.py

```
1 """Implementation of a source-type to display a webcam stream and
       apply the Haar Cascade classifier on it"""
2
3 # Core Pkgs
5 import av
6 import streamlit as st
7 from streamlit_webrtc import webrtc_streamer
9 from pki_a22_app.dashboard.sources import SourcesInterface
10 from pki_a22_app.haarcascades.haarcascades import detect_objects
11
12
13 class WebcamSource(SourcesInterface):
      def load_source(self, classifier_id: str, scale_factor: int,
14
      min_neighbors: int,
                      min_size: int, show_results: bool = False) ->
       None:
16
          This method shows an UI element to display the stream of
      a webcam. The Haar Cascade classifier will then be applied to
       the images
     of the webcam.
```

```
19
          The parameter description was taken from:
20
          https://hackaday.io/project/12384-autofan-automated-
21
      control-of-air-flow/log/41956-face-detection-using-a-haar-
      cascade - classifier
22
23
          Parameters
24
26
          classifier_id : str
              the classifier name. E.g. when you set it to "
      frontalface_default", the configuration file
              "haarcascade_frontalface_default.xml" will be loaded
28
          scale_factor : float, optional, by default 1.1
29
              determines the factor by which the detection window
30
      of the classifier is scaled down per detection pass. A factor
       of 1.1
              corresponds to an increase of 10%. Hence, increasing
31
      the scale factor increases performance, as the number of
      detection
              passes is reduced. However, as a consequence the
      reliability by which a face is detected is reduced. See:
          min_neighbors : int, optional, by default 4
33
              determines the minimum number of neighboring facial
34
      features that need to be present to indicate the detection of
       a face by the
              classifier. Decreasing the factor increases the
35
      amount of false positive detections. Increasing the factor
      might lead to missing
              faces in the image. The argument seems to have no
36
      influence on the performance of the algorithm. See:
          min_size: int
37
              determines the minimum size of the detection window
38
      in pixels. Increasing the minimum detection window increases
      performance.
              However, smaller faces are going to be missed then.
39
      In the scope of this project however a relatively big
      detection window can
              be used, as the user is sitting directly in front of
40
      the camera.
           show_results : bool, optional
41
              If true we want to display the original images AND
42
      the results, by default False
43
          st.subheader("Webcam processing using WebRTC (peer2peer)"
44
45
          def callback(frame):
46
               # convert frame to ndarray
47
              img = frame.to_ndarray(format="bgr24")
49
               # actual object detection
50
51
              result_img, result_faces = detect_objects(
                  img, classifier_id, scale_factor, min_neighbors,
52
      min_size)
53
             return av.VideoFrame.from_ndarray(result_img, format=
```

```
"bgr24")
55
          # This will show the webcam-UI elements in streamlit
56
          # Each frame gets sent to the callback method
57
          rtc_configuration = {  # Add this config
58
               "iceServers": [{"urls": ["stun:stun.1.google.com
59
      :19302"]}]
          }
60
          webrtc_streamer(key="example", video_frame_callback=
      callback,
                           rtc_configuration=rtc_configuration)
```

Listing 7: dashboard/source webcam.py

```
_{\rm 1} """This modules contains helper methods to get information from
      local file system"""
2 import os
3 import pathlib
4 from typing import List
7 def get_classifiers() -> List[str]:
      Reads all files in /resources/haarcascades/ and returns only
9
      the actual classifier
      name without "haarcascade_" and ".xml". "E.g.
10
      haarcascade_frontalface_default.xml"
      gets to "haarcascade_frontalface_default".
11
12
      Returns
13
14
      List[str]
         List of found classifiers
16
17
      classifier_files = os.listdir(os.getcwd() + "/resources/
18
      haarcascades/")
19
      classifiers = [item.replace("haarcascade_", "")
      for item in classifier_files]
classifiers = [item.replace(".xml", "") for item in
20
      classifiersl
      classifiers.sort()
22
      return classifiers
23
24
def get_datasets() -> List[str]:
27
      Returns the folders found under /resources/datasets/
28
29
30
      Returns
31
      List[str]
         List of available dataset
33
34
      datasets = os.listdir(os.getcwd() + "/resources/datasets/")
35
      datasets = [item for item in datasets if not item.startswith(
36
      ".")]
      datasets.sort()
37
    return datasets
```

```
39
40
41 def get_images_of_dataset(dataset_name: str) -> List[str]:
42
       Returns the images contined in a dataset
43
44
45
       Parameters
46
       dataset_name : str
       Name of the dataset (folder)
48
49
      Returns
50
51
      List[str]
52
       List of found images
53
54
      extensions = ['.jpg', '.png', '.jpeg']
img_list = [x for x in pathlib.Path(f"resources/datasets/{
55
56
       dataset_name}").iterdir() if x.suffix.lower() in extensions]
  img_list = [str(item) for item in img_list]
return list(img_list)
57
```

Listing 8: utils/file loader.py