



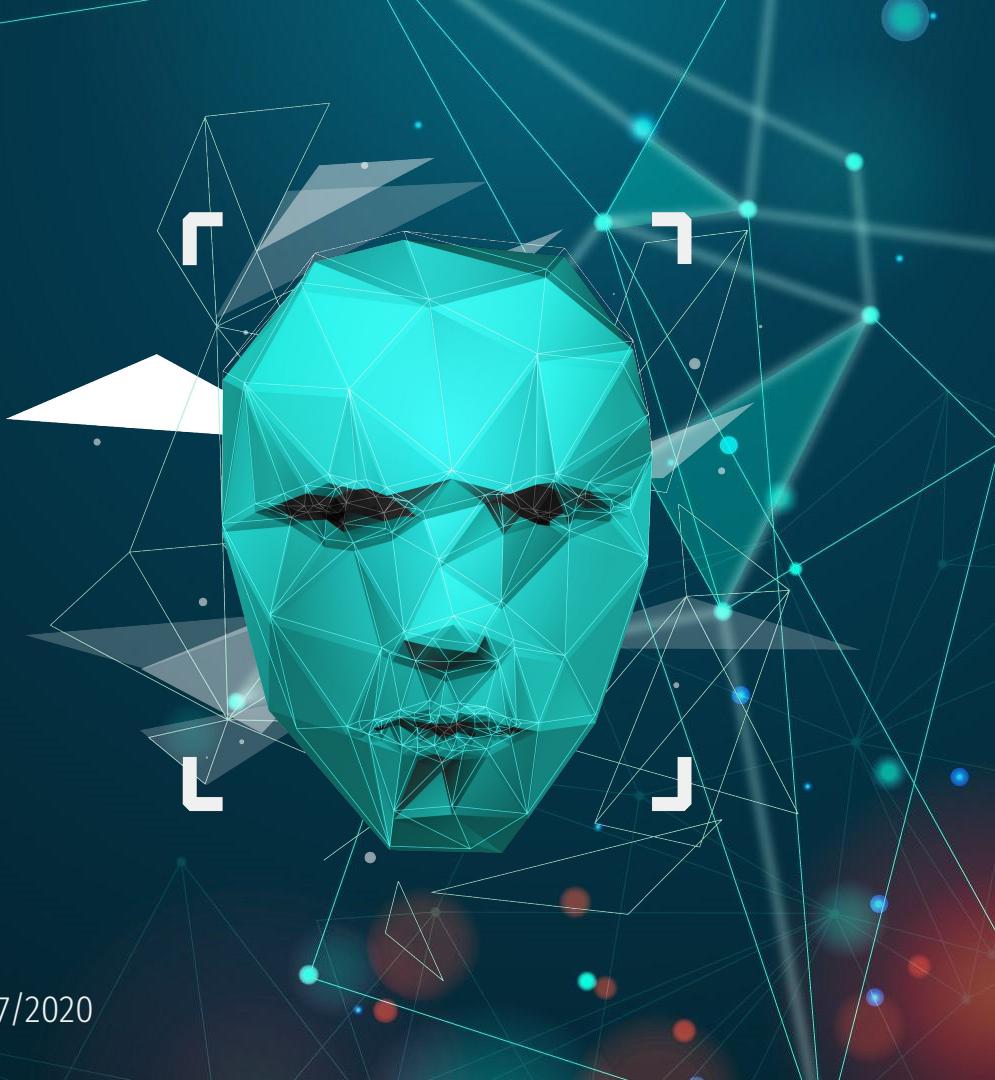
UATTENDED

UAttended

Attendance Checker Based On Face Recognition

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29/07/2020





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Contents

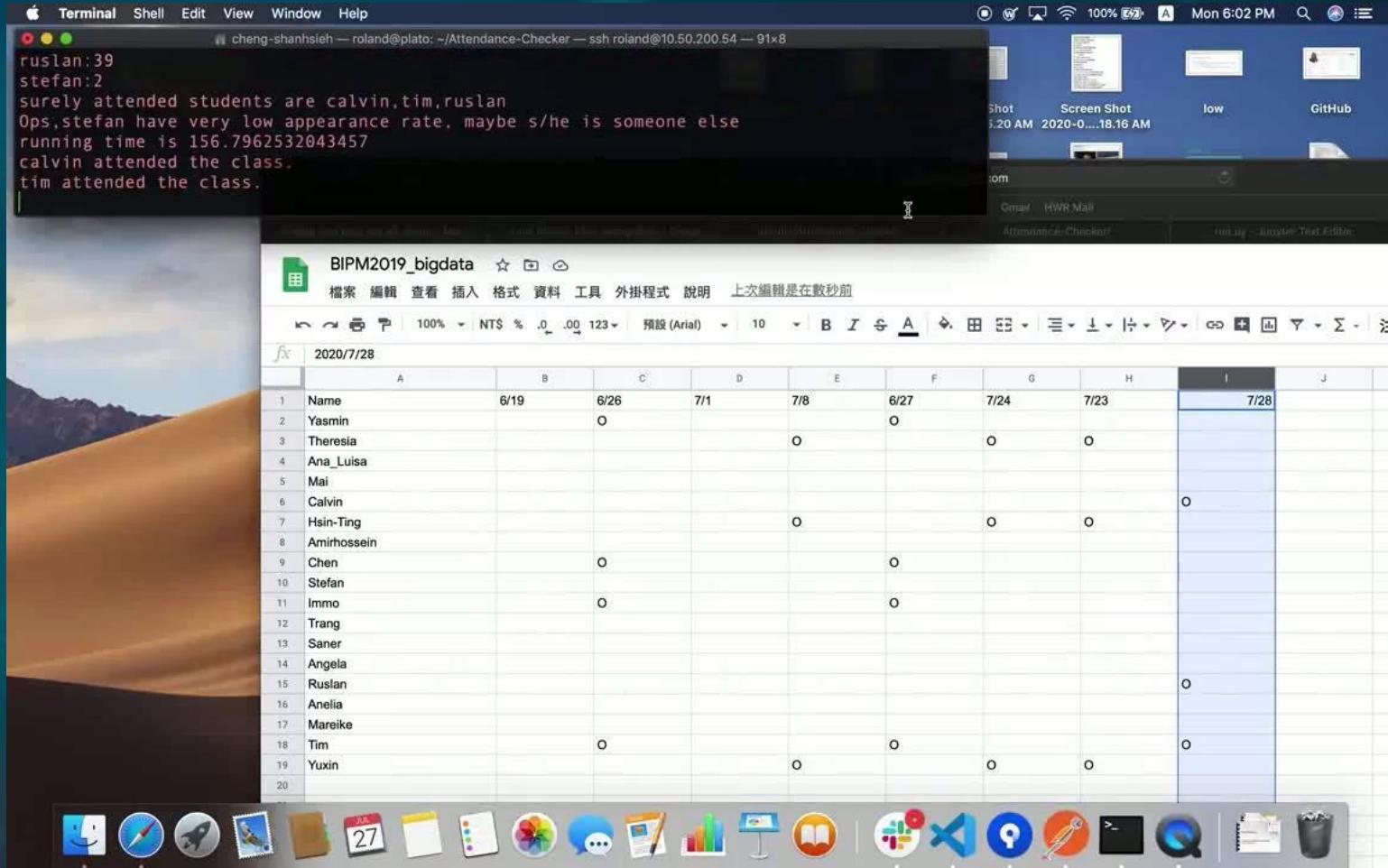
- 01** **Show Time!**
UAttended Demo
The user journey of generating
an attendance list
- 03** **Implementation**
Preparation, training, application,
evaluation, deployment

- 02** **Techniques & Methodology**
Face recognition methods &
tools, GPU, API
- 04** **Conclusion & Outlook**
Summary, reflection & further
work

Show Time! UAttended Demo

01

The user journey of generating an attendance list



Motivation



Attendance Check Is Everywhere

- An important action to evaluate students' dedication, the popularity of lectures/events
- Schools, companies, events all need attendance check

Online Learning Rises Up Dramatically

- Over 1 billion students (pre-primary to tertiary) were affected due to COVID-19
- More and more distant learning programs

Filling-in Attendance List Is Forgettable

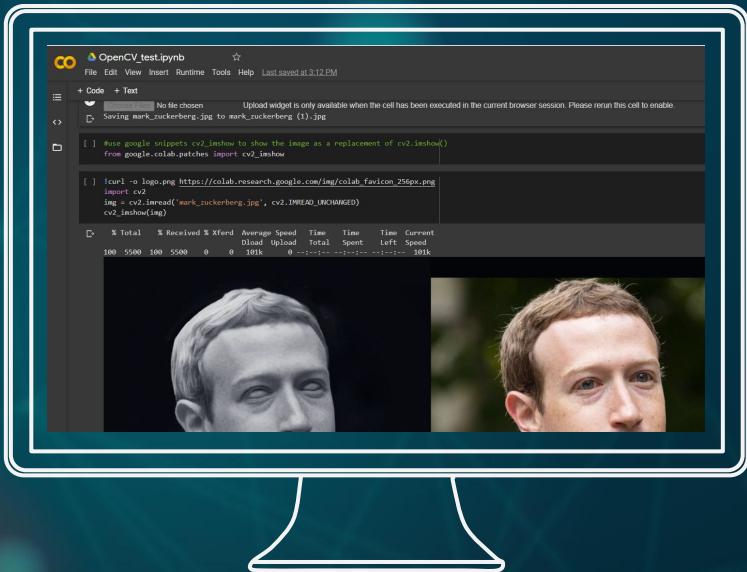
- Check-in in Excel every time for each lecture adds administrative efforts to professors and students
- Too many topics to focus on to remember check-in every time

Techniques & Methodology

02

Face recognition methods & tools,
GPU, API

Working Environment



Google Colaboratory

Development Environment

Language

Python

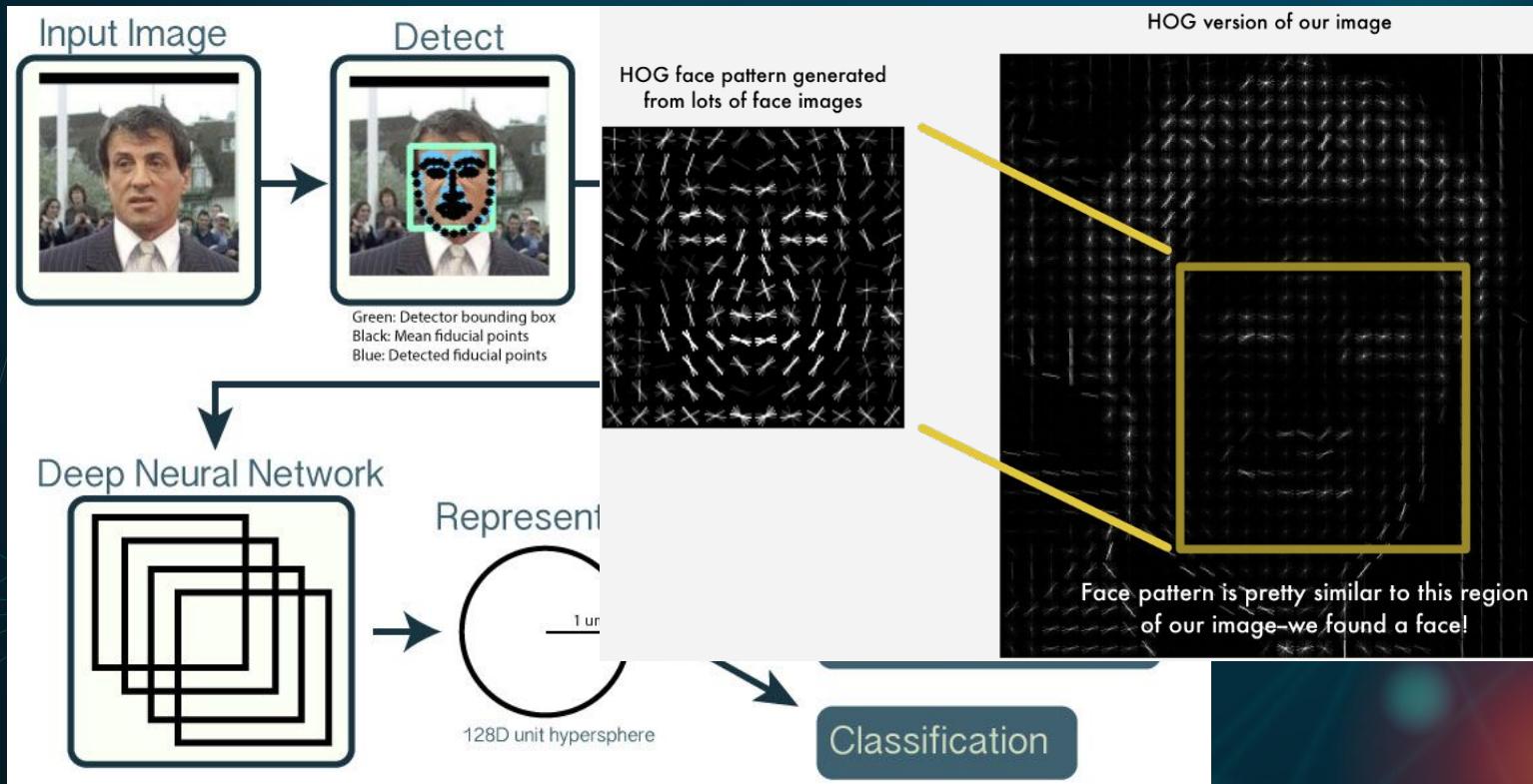
GPU
16GB memory

Processor

Repository

GitHub

Face Recognition



Computer Vision Library - OpenCV

Cascades

- About 6,000 classifiers for a face & thousands of data blocks
- Break the detection into 30 to 50 stages
- Only go to the further stage for detection if the block passes the current stage



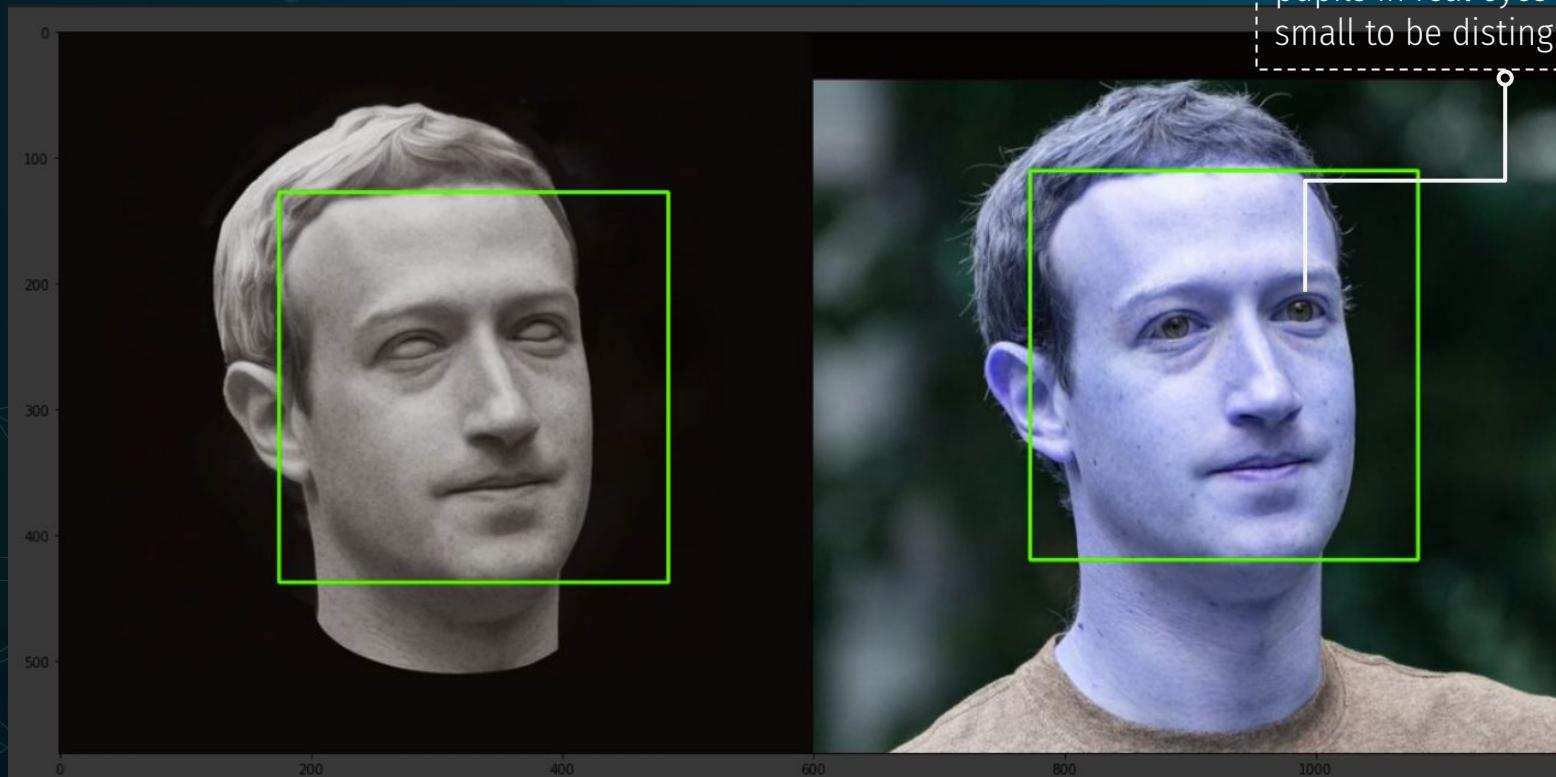
Image Capture & Detection

- Google snippets for `cv2_imshow()` is needed
- Load Cascades xml and apply
- Draw a rectangle around the face

Video Capture

- Via webcam or video files
- capture the video frame by frame.

OpenCV Face Detection Result



Computer Vision Library - OpenCV

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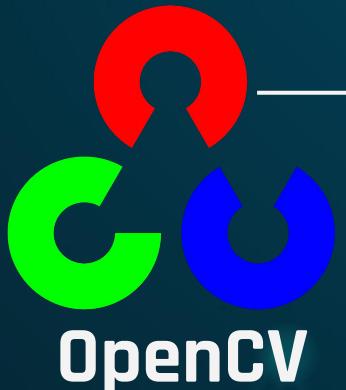


Image Capture & Detection

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Video Capture

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Google Sheets API



- **Google API library (291 public available)**
- **Authentication and authorization**
- **Wide range of functionalities**

```
values = [
    [
        # Cell values ...
    ],
    # Additional rows ...
]
body = {
    'values': values
}
result = service.spreadsheets().values().update(
    spreadsheetId=spreadsheet_id, range=range_name,
    valueInputOption=value_input_option, body=body).execute()
print('{0} cells updated.'.format(result.get('updatedCells')))
```

gspread



- **A Python API for Google Sheets.**
 - Integrate with pandas and Numpy packages

Quick Example

```
import gspread

gc = gspread.service_account()

# Open a sheet from a spreadsheet in one go
wks = gc.open("Where is the money Lebowski?").sheet1

# Update a range of cells using the top left corner address
wks.update('A1', [[1, 2], [3, 4]])

# Or update a single cell
wks.update('B42', "it's down there somewhere, let me take another look.")

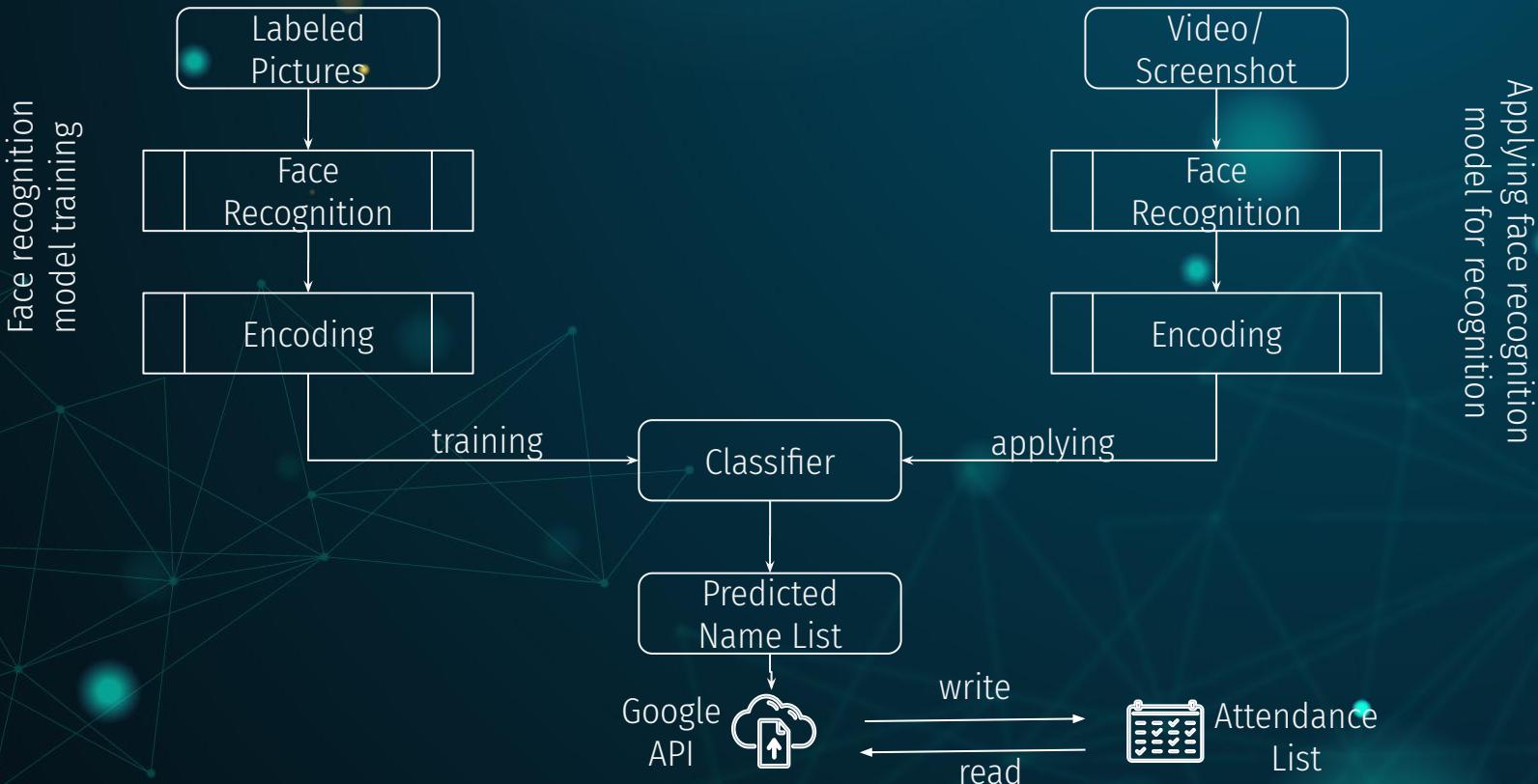
# Format the header
wks.format('A1:B1', {'textFormat': {'bold': True}})
```

Implementation

03

Preparation, training, application,
evaluation, deployment

UAttended Architecture



Preparation



Structure:

```
<test_image>.jpg  
<train_dir>/  
  <person_1>/  
    <person_1_face-1>.jpg  
    <person_1_face-2>.jpg  
    .  
    .  
    <person_1_face-n>.jpg  
  <person_2>/  
    <person_2_face-1>.jpg  
    <person_2_face-2>.jpg  
    .  
    .  
    <person_2_face-n>.jpg  
  .  
  .  
  <person_n>/  
    <person_n_face-1>.jpg  
    <person_n_face-2>.jpg  
    .  
    .  
    <person_n_face-n>.jpg
```

Training Face Recognition Model



Testing Evaluation

01

Testing with Individual Pictures

	precision	recall	f1-score	support
amirhossein	1.00	1.00	1.00	9
ana_luisa	1.00	1.00	1.00	11
anelia	1.00	1.00	1.00	9
angela	1.00	1.00	1.00	5
calvin	1.00	1.00	1.00	8
chen	1.00	1.00	1.00	7
hsin-ting	1.00	1.00	1.00	5
immo	1.00	1.00	1.00	1
mai	1.00	1.00	1.00	3
mareike	1.00	1.00	1.00	10
ruslan	1.00	1.00	1.00	6
saner	1.00	1.00	1.00	1
stefan	1.00	1.00	1.00	3
theresia	1.00	1.00	1.00	5
tim	1.00	1.00	1.00	4
trang	1.00	1.00	1.00	4
yasmin	1.00	1.00	1.00	4
yuxin	1.00	1.00	1.00	8
accuracy	100 %	1.00		103
macro avg	1.00	1.00	1.00	103
weighted avg	1.00	1.00	1.00	103

02

Random Tests with Class Pictures

class (11) true	detected once	class (11) prediction
tim	TRUE	hsin-ting
theresia	TRUE	calvin
ana_luisa	FALSE	ana_luisa
amirhossein	TRUE	theresia
ruslan	FALSE	yuxin
angela	FALSE	mareike
immo	TRUE	chen
chen	TRUE	stefan
mareike	TRUE	ana_luisa
saner	FALSE	immo
hsin-ting	TRUE	tim
stefan	TRUE	mai
yuxin	TRUE	amirhossein
calvin	TRUE	
yasmin	FALSE	
mai	TRUE	
	16 total	13
	4 wrong/missing	
	75,00% accuracy	

Mean accuracy across
four examples

88.84 %

Scenarios

01 Recognize with Screenshot

Pros: Fast and simple

Cons: need to choose the proper moment in case some students are late



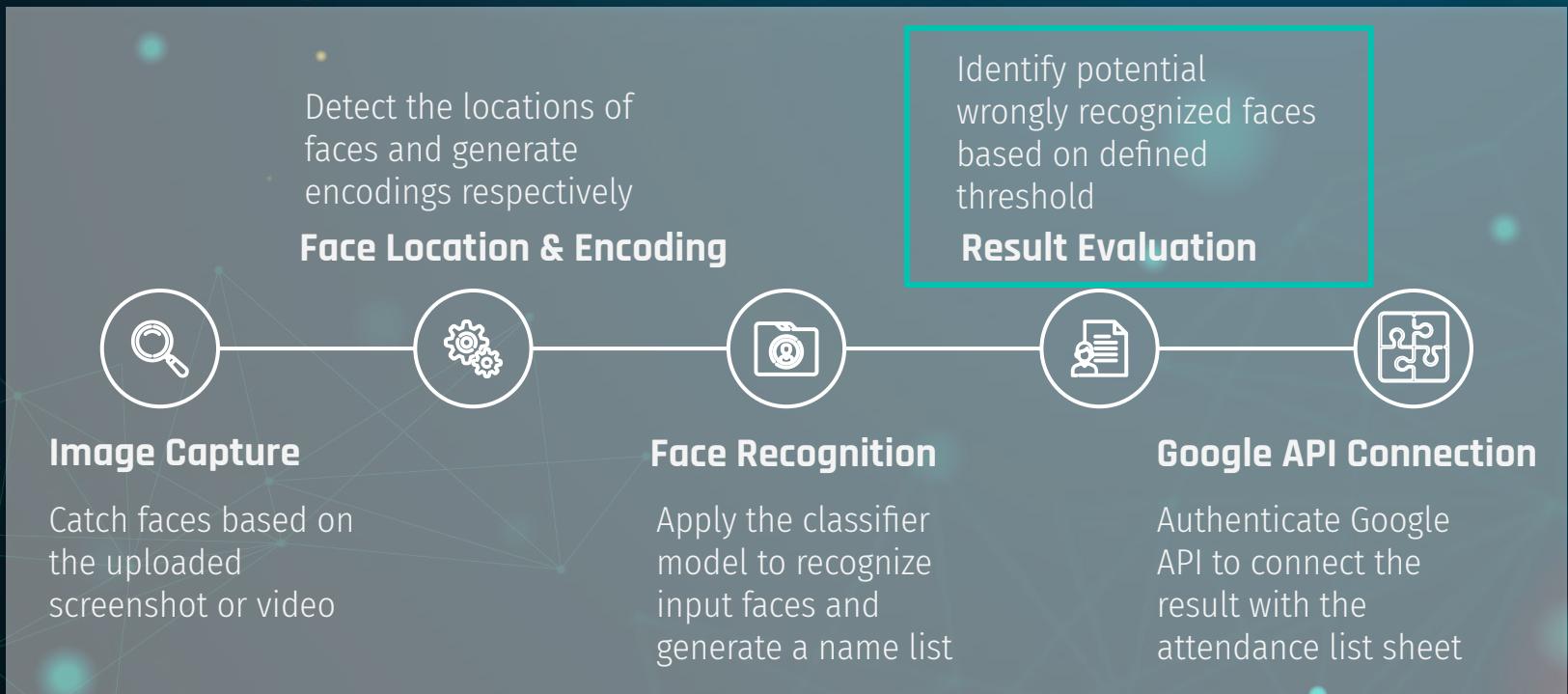
02 Recognize with Videos/Webcam

Pros: No interruption during the lecture, can extract more info, not miss late students

Cons: Complicated computation, privacy concern



Face Recognition Application



Result Evaluation

01

Recognize with Screenshot

identify names
that appear more
than once

Calculate the percentage
of names that could be
wrongly recognized are
calculated

Number of faces detected: 13
students attended are
anelia,saner,calvin,theresia,yuxin,hsin-ting,
ruslan,
angela,chen,immo,amirhossein,tim,stefan.
Ops! Someone looks too similar to
ana_luisa and was wrongly recognized!
7.6% of faces are recognized duplicitely.
running time is 0.008185625076293945.

02

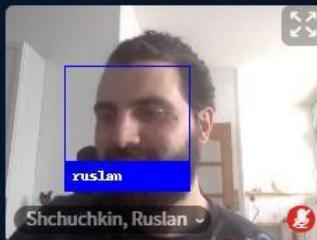
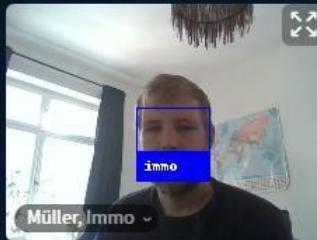
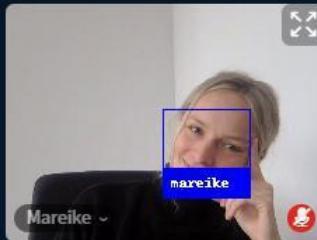
Recognize with Videos/Webcam

Generate a dictionary
{key =name: value =
the frequency
appearing in all
frames}

The frequency of names in all frames are:
yuxin: 63
hsin-ting: 89
theresia 89
ana_luisa: 2
Surely attended students are
yuxin,hsin-ting, theresia.
Ops, ana_luisa has very low appearance
rate, maybe s/he is someone else!

Threshold:
if the frequency of a
name(v_i)/max(v) < 10%
the name could potentially
be wrongly recognized.

Recognition Result



UAttended Deployment on GitHub

hsiehkl / Attendance-Checker

Code Issues Pull requests Actions Projects Wiki Security Insights Settings

master 1 branch 0 tags Go to file Add file Code

hsiehkl Update README.md 827577b 15 minutes ago 25 commits

cred	Add folders	yesterday
material	Add more images for api setup	4 days ago
models	Add folders	yesterday
raw_data	Add folders	yesterday
train_dir	Add folders	yesterday
.gitignore	Update .gitignore	yesterday
README.md	Update README.md	15 minutes ago
api_connector.py	Check empty name list	yesterday
google_sheets.json	Add google_sheets.json example	4 days ago
predict.py	Fix typo	yesterday
run.py	Update run.py	yesterday
train.py	Update train.py	yesterday

About No description, website, or topics provided.

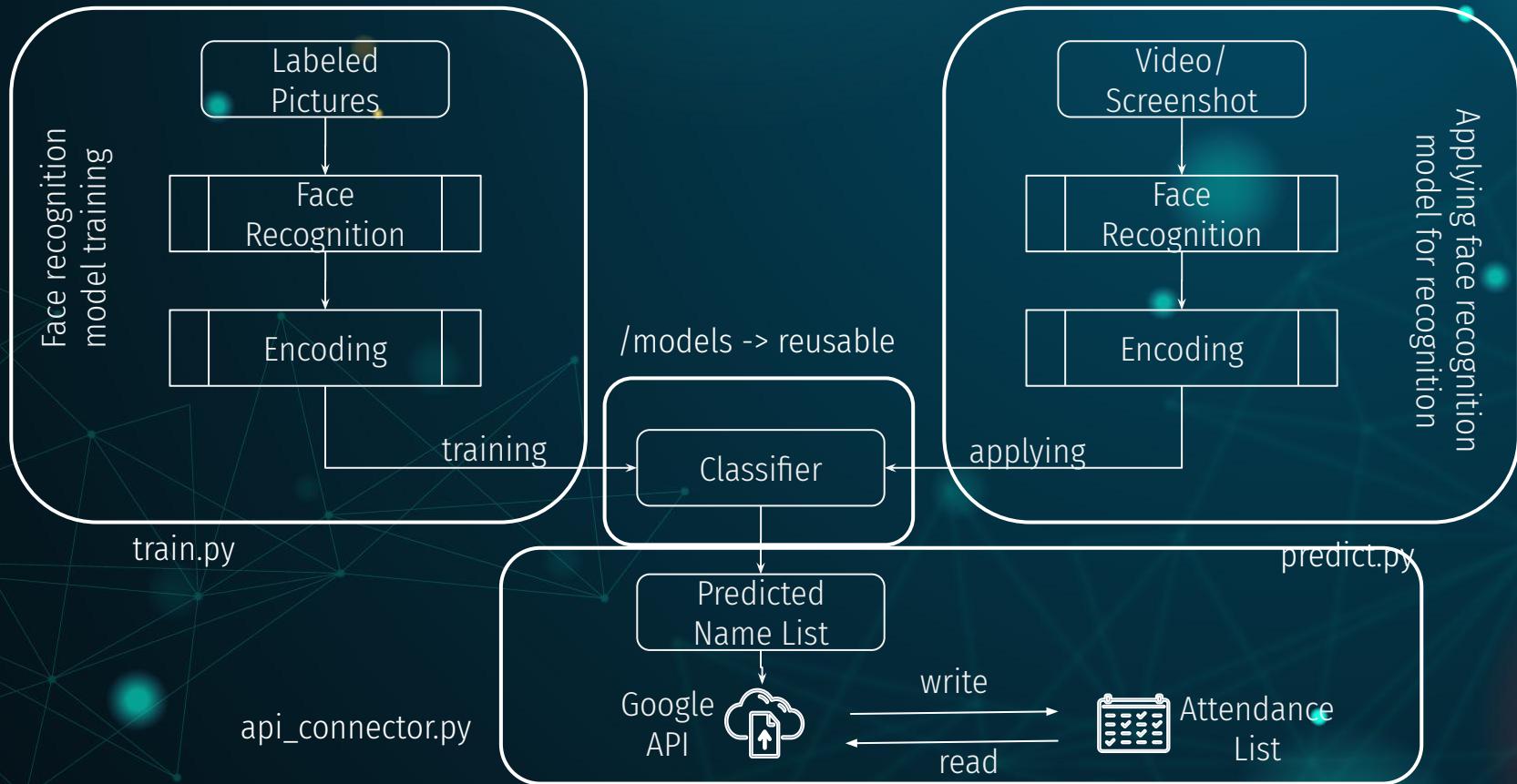
Readme

Releases No releases published Create a new release

Packages No packages published Publish your first package

Languages Python 100.0%

UAttended Main Components



UAttended Deployment on GitHub

<https://github.com/hsiehkl/Attendance-Checker>

Conclusion & Outlook

04

Summary, reflection & further work

**LET AI DO
THE JOB**

Challenges



- **Adaption To Google Colaboratory**

- File reading

- Display the video and image

- **Limited GPU memory**

- Can't reliably apply CNN model

- **<100% Accuracy**

- Fairness

- Subtle manual adjustments needed

- Unknown faces can't be recognized as unknown

- **Data privacy**

- Dissent of students

Misclassification

```
[8] # Load the test image with unknown faces into a numpy array
test_image = face_recognition.load_image_file(data_path_in + '/class/obama.jpg')

# Find all the faces in the test image using the default HOG-based model
face_locations = face_recognition.face_locations(test_image)
no = len(face_locations)
print("Number of faces detected: ", no)

# Predict all the faces in the test image using the trained classifier
name_list = []
print("Found:")
for i in range(no):
    test_image_enc = face_recognition.face_encodings(test_image)[i]
    name = model.predict([test_image_enc])
    name_list.append(*name)
    print(*name)
print(name_list)

⇨ Number of faces detected: 1
Found:
yasmin
['yasmin']
```

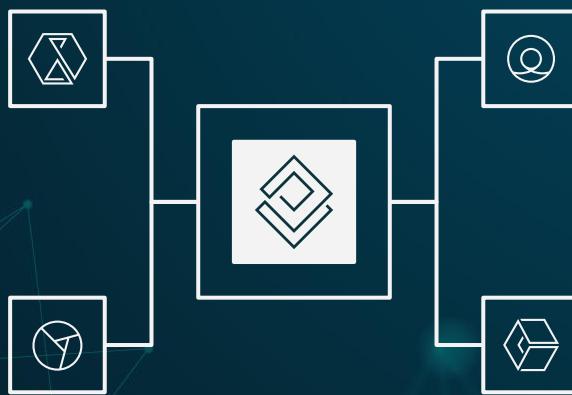


Further Work

Get more powerful GPU
Apply CNN face recognition model
Tune classifier model
Reach 100% accuracy rate

Model Optimization

More Functions
Capture check-in and check-out time
Emotion Analysis
Lecture quality analysis



More user-friendly
Use the service directly without the necessity to download specific packages

Frontend

Privacy Security
Strategy and usage rules to ensure the secure environment and privacy protection

THANKS!

Do you have any questions ?



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Sources

Summary of Face Recognition Steps: Amos, B. (2015, September 24). Summary [Digital image]. Retrieved July 27, 2020, from <https://github.com/cmusatyalab/openface/blob/master/images/summary.jpg>

HOG Face Patterns: [HOG Face Patterns]. (n.d.). Retrieved July 27, 2020, from https://miro.medium.com/max/875/1*6xgev0r-qn4oR88FrW6fiA.png

Folder Structure: Baburajan, A. (2019, December 03). Face_recognition/face_recognition_svm.py at master · ageitgey/face_recognition. Retrieved July 26, 2020, from https://github.com/ageitgey/face_recognition/blob/master/examples/face_recognition_svm.py

Obama: Obama [Digital image]. (n.d.). Retrieved July 26, 2020, from https://raw.githubusercontent.com/ageitgey/face_recognition/master/examples/obama.jpg