CMP304

Project Report

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Human Emotion Recognition

Instructions:

- This is a template that you will fill to complete your assignment report.

- Please read the assessment brief document before attempting this.

- The gray text is meant as guidelines. You are to replace it with your own.

- Delete the instructions part and any gray text before submission.

- After you complete this report, save it as pdf, and submit it along with the compressed folder of your application.

# Introduction

*Relevant overview properly setting the context of the project.*

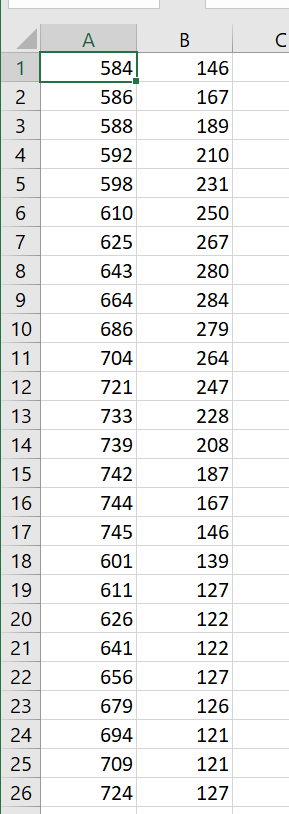
The project is based on the [face\_recognition](https://github.com/ageitgey/face_recognition) python library, one of its python example applications called ‘find\_facial\_features\_in\_picture’ (with my changes to save the results in .csv files) and my own C++ application which uses extracted facial features, saved into .csv files, to learn an emotion and to recognize an emotion from a picture.

# Methodology

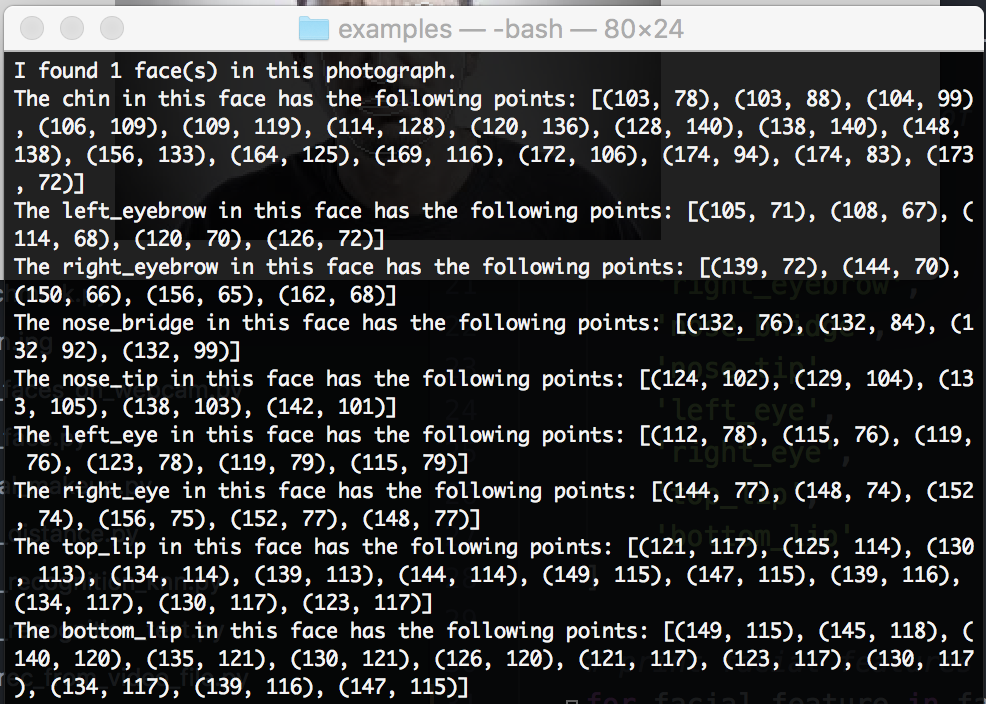
*Description of the steps followed and methods used including a complete explanation and rationale for the techniques and features chosen. You should also acknowledge the tools you used.*

## Python application

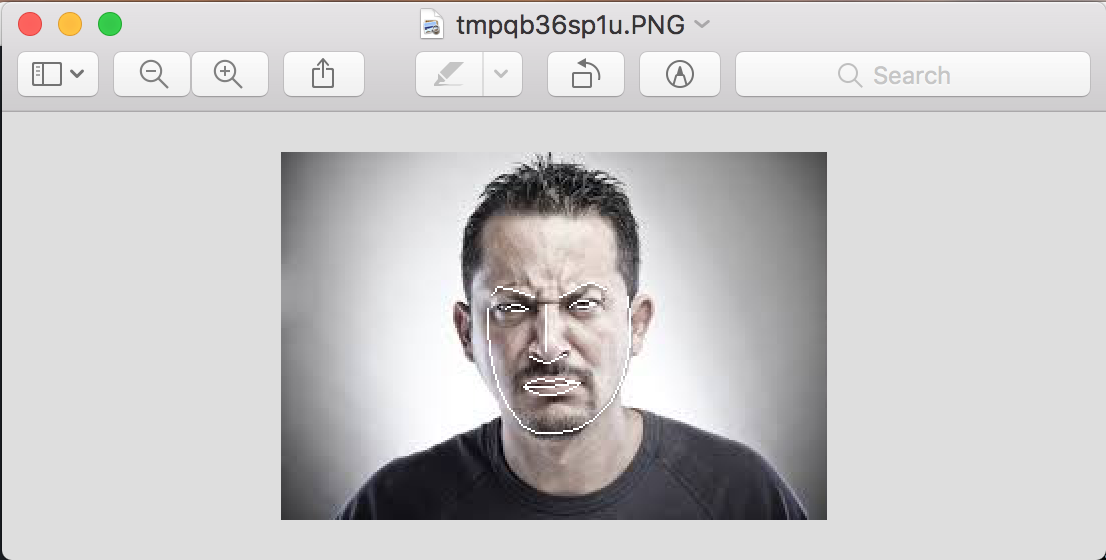
[face\_recognition](https://github.com/ageitgey/face_recognition) python library allows for extracting facial features from a picture. It basically just gives a position (relevant to the picture) of the chin, left eyebrow, right eyebrow, nose bridge, nose tip left eye, right eye, top lip, bottom lip. These positions are saved in a .csv file that look like this:



And this is how the facial features extraction of the python application look like in macOS terminal ([face\_recognition](https://github.com/ageitgey/face_recognition) library works only on Linux and macOS):



Terminal results



Extracted facial features

The python application saves facial features in .csv files that are later used for machine learning of each individual emotion and emotion recognition from a picture, e.g.: 3 smiley faces will be examined by the python application and the results (positions of the extracted facial features) will be saved in one .csv file, e.g.: ‘smile.csv’. Then 4th picture with a smiley face will be examined and the results will be then saved in a .csv file, e.g. ‘smile\_test\_1’.

## C++ application

The C++ application takes the ‘smile.csv’ file which contains all the results for smiley faces, and uses these results to learn from and create weighted results for a ‘smiley’ face (Similarly for ‘angry.csv’, ‘sad.csv’ ect. to learn different emotions). Then ‘smile\_test\_1’ will be examined in the same way (weighted results will be created for ‘smile\_test\_1.csv’) but without specifying what emotion it is. Then the application will compare all the weighted results for different emotions with the weighted results from a picture, that we want to recognize an emotion from, and see which emotion the examined picture fits the best.

## Weighted results

The C++ learning algorithm takes advantage of the fact that after reading the ‘\*.csv’ file into a vector (which is converted into a 1D vector from a 2D .csv file) the size of all facial features form one picture, combined, will be 144 (CHIN = 34, LEFT\_EYEBROW = 10, RIGHT\_EYEBROW = 10, NOSE\_BRIDGE = 8, NOSE\_TIP = 10, LEFT\_EYE = 12, RIGHT\_EYE = 12, TOP\_LIP = 24, BOTTOM\_LIP = 24; += 144). This allows for calculating weighting results for all the pictures in, e.g.: ‘smile.csv’ and then use these results to compare with the picture that we want to recognize an emotion from.

## Weighted results – what is it?

Weighted results are distances between:

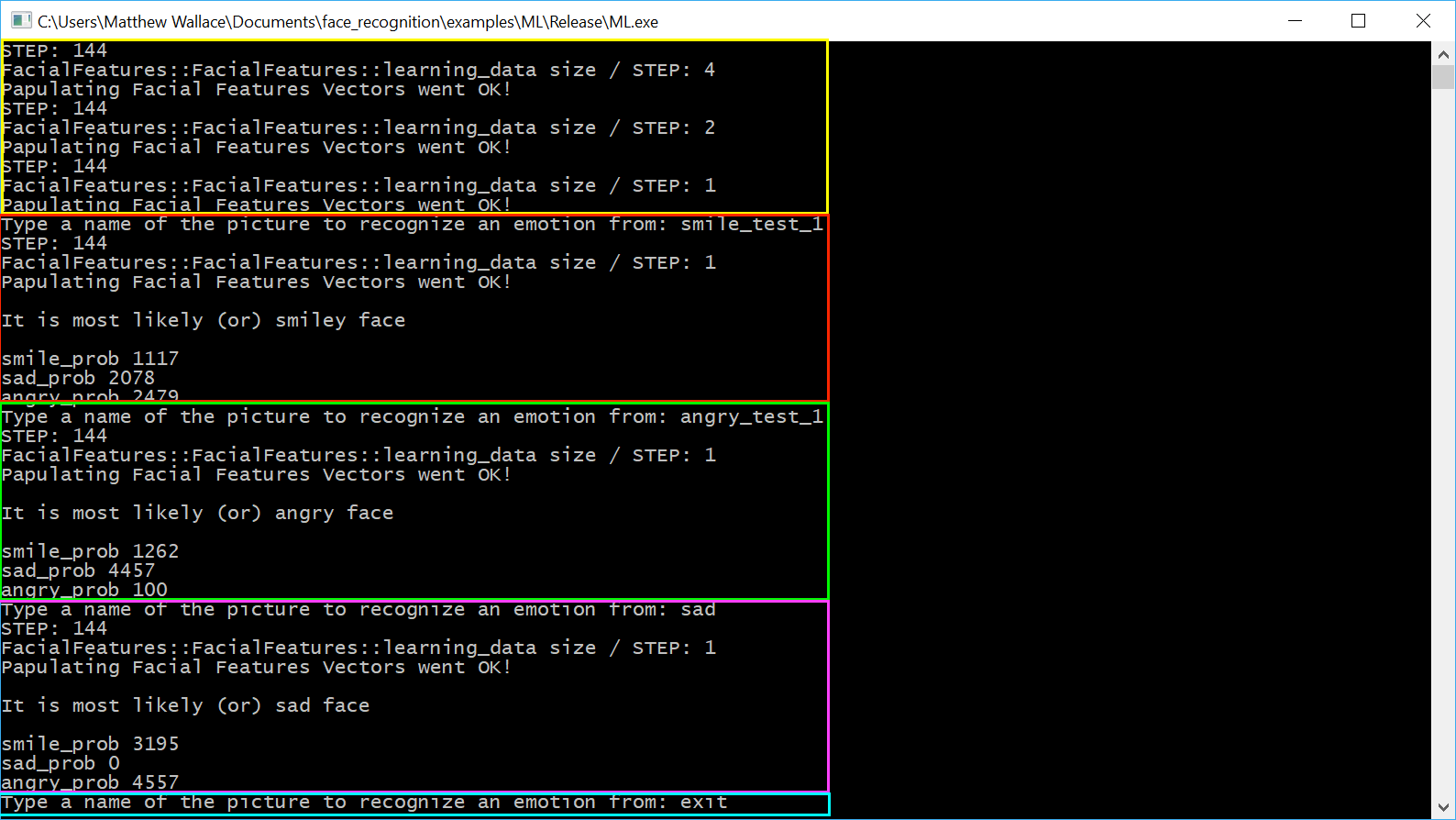
* Left eye and left eyebrow
* Right eye and right eyebrow
* Top lip and bottom lip
* Nose bridge and nose tip.

These differences are calculated to later be stored in a vector (There is a weighted vector produced from a learning file, e.g.: ‘smile.csv’, ‘angry.csv’; and a picture that we want to recognize am emotion from, e.g. ‘smile\_test\_1’, ‘angry\_test\_1’).

Weighted vectors of emotions (smile, angry, sad ect.) are compared with a weighted vector of an emotion that we want to recognize. Because of the method used to create the weighted vectors, the closer the result is to 0, the more likely it is, to be that emotion. This can be really well observed when the learning picture and the picture-to-recognize are the same pictures; then the result will be 0, which is a 100% certainty.

# Results

*Comment on the performance of your application, including test cases. Tabulate and discuss your results. A quantitative measure of performance must be presented.*



Yellow – Learning, Red – smile\_test\_1 picture emotion recognition, Green – angry\_test\_1 picture emotion recognition, Purple – sad picture emotion recognition, Aqua– exit the application

## smile\_test\_1

|  |  |
| --- | --- |
| Probability table | |
| Emotion | Probability |
| Smile | 1117 |
| Sad | 2078 |
| Angry | 2479 |

The probability that the fed image has a smiley face in it, is 1.86 ≈ 2078 ÷ 1117, times bigger than the sad face and 2.22 ≈ 2479 ÷ 1117, times bigger, than the angry face.

The accuracy is 100%, even for test images different that test cases, due to the low number of test cases.

## angry\_test\_1

|  |  |
| --- | --- |
| Probability table | |
| Emotion | Probability |
| Smile | 1262 |
| Sad | 4457 |
| Angry | 100 |

The probability that the fed image has an angry face in it, is 12.62 = 12.62 ÷ 100, times bigger than the smiley face and 44.57 = 4457 ÷ 100, times bigger, than the sad face.

The accuracy is 100%, even for test images different that test cases, due to the low number of test cases.

## Test case - sad

|  |  |
| --- | --- |
| Probability table | |
| Emotion | Probability |
| Smile | 3195 |
| Sad | 0 |
| Angry | 4557 |

The probability that the fed image has a sad face in it, is infinitely bigger than the sad face and the angry face.

The accuracy is 100%, even for test images different that test cases, due to the low number of test cases.

# Conclusions

*Full analysis and summary of the project. Describe techniques used in the project (Supervised learning ect.)*

In the project I used a supervised technique for machine learning.

# References

*A number of references properly cited in Cite Them Right Harvard style.*