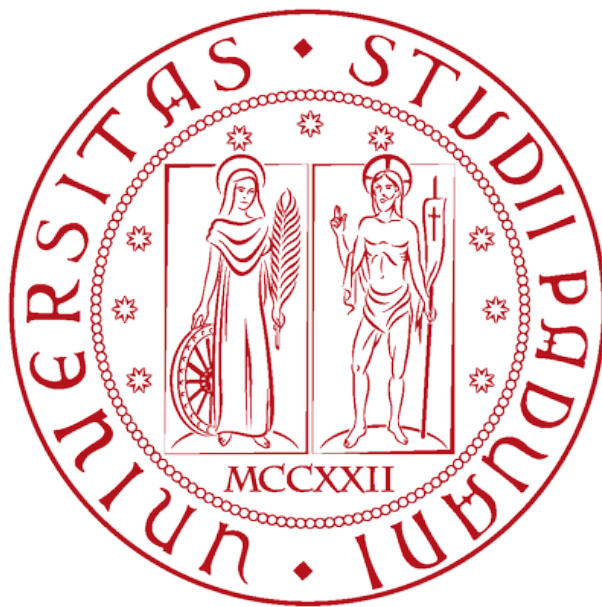


UNIVERSITY OF PADOVA

Computer Vision

Intermediate project report



Marek Tatýrek
Mateusz Mirosław Lis
Abioye Obaloluwa Peter

Academic Year 2024-2025

Contents

1	Introduction	2
1.1	Notes	2
2	Overview	2
3	Implementation details	2
3.1	FileLoader	2
3.2	SceneChangeDetect	3
3.3	TestDatasetEval	3
3.4	FilmStatistics	3
4	Evaluation	4
5	Dataset description	4
6	Results summary	4
7	Individual contributions	4
8	Conclusion	4

1 Introduction

1.1 Notes

To make code more clear, we use:

- `snake_case` for variables
- `PascalCase` for classes
- `camelCase` for functions

2 Overview

We decided to use class design, where classes are logically separated depending on task. Dataflow between classes is provided by several user structures. Data are processed sequentially. Unfortunately used dataflow is not that clear because we haven't been able to fully follow originally defined structure. We also used Doxygen documentation so for detailed description you can check it [here](#).

For loading data we have two classes `ImageLoader()` and `VideoLoader()`, derived from class `InputSource()`. From user side the classes have common interface and usage is the same. Important is method `hasNextFrame()`, used for driving the while loop in the main, returning true in case there is frame or image that we can read. Class is returning timestamp, `cv::Mat` with current sample.

As was already said, tasks are performed sequentially, which is convenient, because we are reading many frames and to store them in the buffer, it will be very memory demanding. Also from the same reason we have to take care about not making deep copies.

We have `Preprocessing()` class, for editing the image before actual detection of haar features.

For detecting features in the images we are using class `HaarDetector()` using haar cascades for finding desired patterns in the images. In our case faces and eyes.

From the detected features we need to make evaluation and decide which shot type are we having. For this purpose we have class `FeatureEvaluator()` that outputs structure `classification_result` with information if it is current sample wide shot, medium or close up.

Because we want to make statistic from the data, we made `FilmStatistics()` class. It makes sense to use this class only on video data. At the init we provide configuration structure `FilmStatisticsEvalConfig`, with many settings. We can export time sequences to `.csv` file or we can use getters to get the time sequences and use them in the code.

For graphical output of the statistic data we have `ResultDisplay()` class, which is inputting all data types got from `FilmStatistics()` getters and returning `cv::mat` with plots.

3 Implementation details

3.1 FileLoader

In this file we have four classes.

Class `InputSource` is parent class for `ImageLoader` and `VideoLoader`. It serves to provide easy interface to work with image and video files. In both cases we are aiming to process data in series, that means frame by frame. Both of these classes have same user interface made of:

- `hasNextFrame()` - returns if there is one more frame to process, used for driving while loop.
- `nextFrame()` - returns next frame in the sequence.
- `getCurrentTimestamp()` - returns current timestamp in ms.

For preprocessing we have class `Preprocessing()`. We only implemented processing methods that we found beneficial. The class contains this methods:

- `LoadFrame()`
- `resizeImage()`
- `toGrayscale()`
- `equalizeHistogram()`
- `GetProcessedImage()`

Names on methods are describing function sufficiently.

3.2 SceneChangeDetect

This file contains single function for optimizing processing of Viola & Johnes detector. As an input we take two frames, we compute how much they are different and return boolean value. Because Viola & Johnes is computationally quite expansive, we want to process the frame by it, only when it is really needed.

We have two parameters:

- `pixelDiffThreshold` - Difference between two pixel at same position in input frames.
- `threshold` - Percentage of pixels, that reach `pixelDiffThreshold` parameter value.

Algorithm computes difference between all pixel at appropriate positions and make boolean array. Then it computes percentage of pixels that reaches the `pixelDiffThreshold` and compare the value with `threshold`. If the computed value reaches the threshold it returns true.

3.3 TestDatasetEval

This simple class is designed for computing accuracy of the algorithm on the test dataset. That means at image dataset. Since we have test dataset sorted in the folders, task is pretty simple. We have class `TestDatasetEval`. with constructor we pass `ShotType`, which we want to test with this class. Then we have two methods:

- `check()` Is used in while to add currently classified image `ShotType` to evaluation.
- `getEvalResults()` After all images are processed we call this method and it returns accuracy.

3.4 FilmStatistics

Class is mainly designed for video. As input it takes class probabilities and it outputs .csv file with statistical data. Constructor takes one argument - structure `FilmStatisticsEvalConfig`, which is used to set up all the parameters for class.

We have 4 public methods:

- `addConfigurationStruct()` - Serves to change configuration settings.
- `addFrameResult()` - Add frame to final evaluation.
- `exportToCSV()` - Exports computed statistical timelines to csv file for future processing.
- `printSummary()` - Prints basic statistical summary.

It is important to mention what specifically class does. It takes input probability. We can set oversampling or skipping to input data to filter random noise etc. Class computes entropy and entropy variance, it detects cuts in the scenes and classify most probable `ShotType`. We can set different window sizes for oversampling window, entropy window and entropy variance window. We can also reduce output data flow if we dont want to have informations from each single shot and we reduce size of output csv file. All of these parameters are settable in `FilmStatisticsEvalConfig`.

Since we are using sliding windows for signal processing, we introduce delay to the system, which grows as we setting bigger window sizes. Delay is different for each statistical parameter timeseries. Class handle this situation and starts outputing data when all of the parameters are valid, with correct timestamp.

4 Evaluation

5 Dataset description

6 Results summary

7 Individual contributions

8 Conclusion

References