

IMPLEMENTATION OF ALGORITHMS FOR THE RESTORATION OF RETINAL IMAGES

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**LA SALLE UNIVERSITY
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Degree work to obtain the title of engineer in automation

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Thesis approval

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Contents

Introduction	11
Glossary	12
data persistence	12
De-serialization	12
Serialization	12
Level	12
Channel	12
Histogram	12
Intensity	12
BGR image	12
BGRA image	13
Descriptor	13
Threshold	13
Segmentation	13
Mask	13
Perspective transform	13
Filter	13
hashable	13
hashing	13
Numpy Arrays	13
ndarray.ndim	13
ndarray.size	13
ndarray.shape	13
ndarray.dtype	13
Art State	14
Programming and Research Approach	15
Optimization Methods	15
Process big data as small data	16
Lazy evaluation	16
Preventing circular references	16
Multitasking	16
Memory mapped files	16
Memoization and caching	16
Images in openCV	16
Algorithms design	18
General algorithms	18
The area under a polygon (<i>polygonArea</i>)	18
Overlay	20
Real image – Rendered image	20
Domain transformations: scaling to original results	21
Normalization	23
Load functions	23
Load from sockets	23
Load from URLs	23
Load from files	23
Load from memmapped files	23
Pre-processing, Filters and Enhancing Methods	23
Histogram equalization	23

Matrix decomposition to reduce levels	24
Smoothing with 1D-filters	25
Gaussian Filter	27
Bilateral Filter	27
SigmoidImageFilter	29
Normalized SigmoidImageFilter	29
Custom filters using <i>normSigmoid</i>	31
Sigmoid filtering and saturation	34
Segmentations	34
Convex hull with line cuts	34
Binary masks	36
Alpha masks	37
Object Recognition and Matching algorithms	39
Entropy	39
Histogram comparison	39
Histogram matching	39
Detecting and computing features	40
ASIFT	40
Using rates and probabilities	40
Convexity ratio	41
Rectangularity	43
Implementation	47
User Interface	74
Command-line interface	74
Executable	75
Results	75
Result for set 1	79
Result for set 2	80
Result for set 3	82
Result for set 4	83
Result for set 5	84
Result for set 6	85
Result for set 7	86
Result for set 8	87
Result for set 9	89
Result for set 10	91
Result for set 11	92
Result for set 12	93
Result for set 13	94
Result for set 14	95
Result for set 15	96
Result for set 16	98
Result for set 17	100
Result for set 18	102
Result for set 19	104
Result for set 20	108
Result for set 21	111
Result for set 22	113
Result for set 23	114
Result for set 24	116
Result for set 25	117

Result for set 26	119
Result for set 27	120
Validation	120
Discussion	120
Conclusions	121
Recommendations	121
Future Work	122
References	123
Appendix A RRtoolbox: Python Implementation	125
A.1 Procedures to set-up <i>RRtoolbox</i>	125
A.1.1 Install Python	126
A.1.2 Install pip	126
A.1.3 Install packages to Python using pip	126
0.2 Procedures to generate <i>RRtoolbox</i> documentation	128
0.3 Procedures to download <i>inrestore</i> program and source	128

List of Figures

1 "4+1" View Model	11
2 Used representations	17
3 Axes orientation of images	17
4 File estructure of <i>RRtoolbox</i> package	19
5 Find the area of a polygon using Equation 1	20
6 Overlay example using Equation 2	21
7 Histogram equalization example	25
8 CLAHE	26
9 Decomposition tests	27
10 Savgol filter	27
11 Filter with Hanning window	28
12 Filter with Hanning window and shifted convolution	28
13 Spacial filters comparison	29
14 Selection of bilateral parameters according to image shape	30
15 Bilateral filter exposed to different shapes	31
16 Common filters response using <i>normSigmoid</i>	32
17 Common filters comparison	33
18 Filters class diagrams	33
19 Custom filtering without saturation	34
20 Ideal threshold with defect lines	35
21 Practical threshold with defect lines	35
22 Histogram analysis to find threshold values to use in watershed method	36
23 Brightest areas in image using watershed method	37
24 Spacial filters comparison	38
25 Alpha mask histogram analysis and filters response	40
26 Alpha mask obtained using filters	41
27 Alpha mask obtained using layered masks	42
28 Overlay example using alpha masks	42

29	Ordered images using entropy	43
30	Ordered images using histogram comparison	44
31	Matching example	45
32	Features using SIFT	45
33	Convexity ratio of an object	46
34	Rectangularity example	46
35	<i>imrestore.py</i> classes: <i>ImRestore</i> and its inheritor <i>RetinalRestore</i>	48
36	Persistence to <i>descriptors</i> folder and its manipulation	76
37	Images in set1	79
38	Result for set1 using command ‘imrestore ../results/set1/*.* --lens --overwrite --cachePath {temp}’	79
39	Result for set2 using command ‘imrestore ../results/set2/*.* --lens --overwrite --cachePath {temp}’	80
40	Images in set2	80
41	Customized result for set2 using command ‘imrestore ../results/set2/*.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}’	81
42	Result for set3 using command ‘imrestore ../results/set3/*.* --lens --overwrite --cachePath {temp}’	82
43	Images not used in restoration of set3	82
44	Result for set4 using command ‘imrestore ../results/set4/*.* --lens --overwrite --cachePath {temp}’	83
45	Images in set4	84
46	Result for set5 using command ‘imrestore ../results/set5/*.* --lens --overwrite --cachePath {temp}’	85
47	Images not used in restoration of set5	85
48	Result for set6 using command ‘imrestore ../results/set6/*.* --lens --overwrite --cachePath {temp}’	86
49	Images in set6	86
50	Result for set7 using command ‘imrestore ../results/set7/*.* --lens --overwrite --cachePath {temp}’	87
51	Images not used in restoration of set7	87
52	Result for set8 using command ‘imrestore ../results/set8/*.* --lens --overwrite --cachePath {temp}’	88
53	Images in set8	88
54	Result for set9 using command ‘imrestore ../results/set9/*.* --lens --overwrite --cachePath {temp}’	89
55	Images in set9	89
56	Result for set10 using command ‘imrestore ../results/set10/*.* --lens --overwrite --cachePath {temp}’	91
57	Images in set10	91
58	Result for set11 using command ‘imrestore ../results/set11/*.* --lens --overwrite --cachePath {temp}’	92
59	Images in set11	92
60	Result for set12 using command ‘imrestore ../results/set12/*.* --lens --overwrite --cachePath {temp}’	93
61	Images in set12	93
62	Result for set13 using command ‘imrestore ../results/set13/*.* --lens --overwrite --cachePath {temp}’	94
63	Result for set14 using command ‘imrestore ../results/set14/*.* --lens --overwrite --cachePath {temp}’	95
64	Images in set14	95
65	Result for set15 using command ‘imrestore ../results/set15/*.* --lens --overwrite --cachePath {temp}’	97
66	Images not used in restoration of set15	97
67	Result for set16 using command ‘imrestore ../results/set16/*.* --lens --overwrite --cachePath {temp}’	98
68	Result for set17 using command ‘imrestore ../results/set17/*.* --lens --overwrite --cachePath {temp}’	101
69	Images not used in restoration of set17	101
70	Result for set18 using command ‘imrestore ../results/set18/*.* --lens --overwrite --cachePath {temp}’	102
71	Images in set18	103

72	Result for set19 using command ‘imrestore ..//results/set19/*.* --lens --overwrite --cachePath {temp}’	106
73	Images not used in restoration of set19	107
74	Result for set20 using command ‘imrestore ..//results/set20/*.* --lens --overwrite --cachePath {temp}’	108
75	Images in set20	110
76	Result for set21 using command ‘imrestore ..//results/set21/*.* --lens --overwrite --cachePath {temp}’	111
77	Images in set21	111
78	Result for set22 using command ‘imrestore ..//results/set22/*.* --lens --overwrite --cachePath {temp}’	113
79	Images not used in restoration of set22	113
80	Result for set23 using command ‘imrestore ..//results/set23/*.* --lens --overwrite --cachePath {temp}’	114
81	Images in set23	115
82	Result for set24 using command ‘imrestore ..//results/set24/*.* --lens --overwrite --cachePath {temp}’	116
83	Images in set24	116
84	Result for set25 using command ‘imrestore ..//results/set25/*.* --lens --overwrite --cachePath {temp}’	117
85	Images not used in restoration of set25	117
86	Result for set26 using command ‘imrestore ..//results/set26/*.* --lens --overwrite --cachePath {temp}’	119
87	Images in set26	119
88	Result for set27 using command ‘imrestore ..//results/set27/*.* --lens --overwrite --cachePath {temp}’	120
89	<i>RRtoolFC</i> Main developed Graphical Interface	122
90	<i>RRtoolFC</i> ’s console	123
91	Functional application of the <i>RRtoolFC</i> ’s console	123
92	Finding release link in <i>RRtools</i> repository	128
93	<i>Imrestore</i> pre-release	129
94	<i>Imrestore</i> downloaded sources	129

List of Tables

1	Comparison of programming languages	15
2	Comparison example of "for loops" in C and Python	15
3	Output messages for first execution of python imrestore.py tests/im1* --lens --overwrite --cachePath . command	77
4	Output messages for second execution of python imrestore.py tests/im1* --lens --overwrite --cachePath . command	78
5	Output messages for execution of python imrestore.py tests/im1* --lens --overwrite --cachePath . command with modified cache	78
6	Profiling for set1 using command ‘imrestore ..//results/set1/*.* --lens --overwrite --cachePath {temp}’	79
7	Profiling for set2 using command ‘imrestore ..//results/set2/*.* --lens --overwrite --cachePath {temp}’	80
8	Profiling for set2 using command ‘imrestore ..//results/set2/*.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}’	81
9	Profiling for set3 using command ‘imrestore ..//results/set3/*.* --lens --overwrite --cachePath {temp}’	82

10	Profiling for set4 using command ‘imrestore ../results/set4/*.* --lens --overwrite --cachePath {temp}’	83
11	Profiling for set5 using command ‘imrestore ../results/set5/*.* --lens --overwrite --cachePath {temp}’	84
12	Profiling for set6 using command ‘imrestore ../results/set6/*.* --lens --overwrite --cachePath {temp}’	85
13	Profiling for set7 using command ‘imrestore ../results/set7/*.* --lens --overwrite --cachePath {temp}’	86
14	Profiling for set8 using command ‘imrestore ../results/set8/*.* --lens --overwrite --cachePath {temp}’	87
15	Profiling for set9 using command ‘imrestore ../results/set9/*.* --lens --overwrite --cachePath {temp}’	89
16	Profiling for set10 using command ‘imrestore ../results/set10/*.* --lens --overwrite --cachePath {temp}’	91
17	Profiling for set11 using command ‘imrestore ../results/set11/*.* --lens --overwrite --cachePath {temp}’	92
18	Profiling for set12 using command ‘imrestore ../results/set12/*.* --lens --overwrite --cachePath {temp}’	93
19	Profiling for set13 using command ‘imrestore ../results/set13/*.* --lens --overwrite --cachePath {temp}’	94
20	Profiling for set14 using command ‘imrestore ../results/set14/*.* --lens --overwrite --cachePath {temp}’	95
21	Profiling for set15 using command ‘imrestore ../results/set15/*.* --lens --overwrite --cachePath {temp}’	96
22	Profiling for set16 using command ‘imrestore ../results/set16/*.* --lens --overwrite --cachePath {temp}’	99
23	Profiling for set17 using command ‘imrestore ../results/set17/*.* --lens --overwrite --cachePath {temp}’	100
24	Profiling for set18 using command ‘imrestore ../results/set18/*.* --lens --overwrite --cachePath {temp}’	102
25	Profiling for set19 using command ‘imrestore ../results/set19/*.* --lens --overwrite --cachePath {temp}’	105
26	Profiling for set20 using command ‘imrestore ../results/set20/*.* --lens --overwrite --cachePath {temp}’	108
27	Profiling for set21 using command ‘imrestore ../results/set21/*.* --lens --overwrite --cachePath {temp}’	111
28	Profiling for set22 using command ‘imrestore ../results/set22/*.* --lens --overwrite --cachePath {temp}’	113
29	Profiling for set23 using command ‘imrestore ../results/set23/*.* --lens --overwrite --cachePath {temp}’	114
30	Profiling for set24 using command ‘imrestore ../results/set24/*.* --lens --overwrite --cachePath {temp}’	116
31	Profiling for set25 using command ‘imrestore ../results/set25/*.* --lens --overwrite --cachePath {temp}’	117
32	Profiling for set26 using command ‘imrestore ../results/set26/*.* --lens --overwrite --cachePath {temp}’	119
33	Profiling for set27 using command ‘imrestore ../results/set27/*.* --lens --overwrite --cachePath {temp}’	120

List of Equations

1	Area under a polygon	18
2	Overlay images	20
3	Real to rendered	21

4	Rendered to real	21
5	Original TM from scaled TM	22
6	Normalization	23
7	SigmoidImageFilter	29
8	Normalized sigmoid	29
9	Sigmoid function	30
10	Lowpass filter	31
11	Highpass filter	31
12	Bandstop filter	31
13	Bandpass filter	32
14	Inverted Bandstop filter	32
15	Inverted Bandpass filter	32
16	Convexity Ratio	41
17	Rectangularity Ratio	43

List of Algorithms

List of Codes

1	Histogram equalization MWE	24
2	Decomposition MWE	26
3	MWE to create alpha mask with filters	39
4	Imrestore's script code	49
5	General command to process the retinal sets	76
6	Install OpenCV	127

Introduction

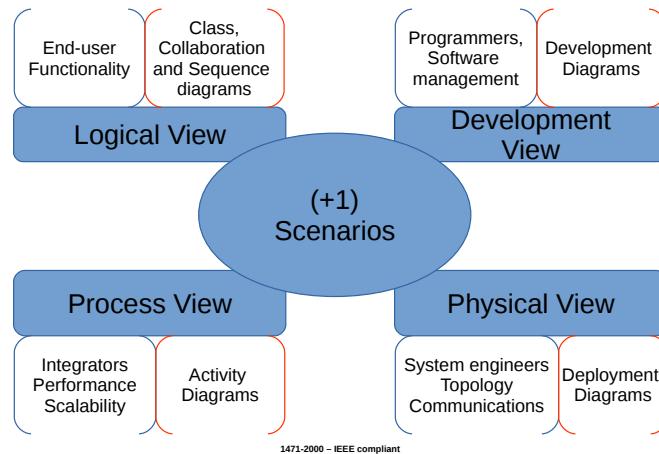
This document was conceived as a phase of a project at the AVARC research group of the Optometry Faculty at La Salle University. The phase consists in the implementation of an algorithm for the restoration of retinal images taken by mobile devices so that it can be further analysed and diagnosed to identify potential patients suffering diabetic retinopathy. It was taken into account the following objectives to ensure the implementation of the application:

1. **Develop algorithm(s) for image restoration:** In which a set of tools (algorithms) were developed for the realization of partial goals dividing the problem into several stages and processing each stage's result to get near the desired application.
2. **Quantitatively validate the application of the restoration algorithm(s):** Each algorithm was validated with its theory and in some cases, a comparison was carried out between its estimation and the desired output.
3. **Implement a restoration application for images:** All the tools were organized and packed into one module according to their category and with some of them a program was developed to offer an automated solution for the restoration of retinal images.

In the world of computer science, some old methods remain robust and effective to solve their intended problems, so in this context some archaic algorithms were included in the main distribution of the program and were not redesigned, as a saying says "there is no need to reinvent the wheel" and thus these algorithms were wrapped into tools. Some algorithms are named in this document but not demonstrated either by its simplicity or because they are thoroughly documented and can be found easily in other sources. All the algorithms are documented and the main ones are tested by means of demonstrations, comparisons or using "unittest" (i.e. Unit testing framework) to ensure robustness and reliability of the code.

Generally the "4+1" View Model (see Figure 1) is widely used to explain complex systems or to convey program models (Phillippe Kruchten, 1995) not only because it complies to the IEEE standard (IEEE, 2000) but because of its simplicity. Despite that, to ensure a clear explanation and understanding of the algorithms this document uses mainly pseudo-codes with some concepts while leaving the code implementation out in the annexes for demonstration purposes. So the "4+1" View Model is partially used here showing in some cases the use classes, some diagrams and leaving out other models. As a plus, additional non-standard tests and demonstrations are given for the reader to confirm that each algorithm works and can be used in practical examples.

Figure 1. "4+1" View Model



Due to that it is not the main purpose of this study to show the complete functionalities of the algorithms but just their implementation some of them are explained using pseudo-code (used to code to actual algorithm or to abstract long ones), others

implemented in their language (to emphasize details and functionality) and in some cases other resources (e.g. explanation, class diagrams, outputs, etc.). In this way, codes that explains themselves, are too short or depend heavily in the programming language definitions are briefly discussed with a Minimal Working Example (MWE) or their pseudo-code not necessarily corresponding to the implemented one. More information is in the appendix and the reader is invited to review the external sources for completeness and the algorithms not explained in this document.

Glossary

The concepts used in this document are a bit esoteric due to the vast majority of specialized procedures used in the algorithms but the intent of this is to make them as simple as possible to let the reader understand the document more easily. Below is a list of key words used throughout the document to encapsulate special concepts that may clarify and convey a better understanding of their meanings:

data persistence

In the context of a program, it consists in storing processed data in a “persistent” form to a non-volatile storage so that the next time the program is run it does not process the data again. This is useful for information that is frequently accessed but that is difficult to obtain e.g. Objects can be stored as byte streams (serialized) and then recreated back (de-serialized) when needed in a program.

De-serialization

It is the contrary process of serialization.

Serialization

It is the process of converting an object into a stream of bytes in order to store the object or transmit it to memory, a database, or a file. Its main purpose is to save the state of an object in order to be able to recreate it when needed.

Level

It represents a value that any pixel can take on an image i.e. there are usually 256 levels in an gray image and 2 levels in a binary image.

Channel

It refers to the color band in an image i.e. BGRA images have 4 channels compared to the single channel in GRAY images.

Histogram

Graphical representation (or its data equivalent) of the intensity distribution or the number of pixels in each level (intensity value) of a channel in an image.

Intensity

Pixel-level or the pixel’s value.

BGR image

Refers to a RGB image with its channels reversed, thus channel 0 is for blue color conversely to Red in RGB images.

BGRA image

It is an RGB image that contains an additional A channel to support additional information like transparency for PNG formats.

Descriptor

Is the description of a feature in an image where elementary characteristics such as shape, color, texture, etc are taken to describe the feature.

Threshold

It is a type of image segmentation at which an image is partitioned into a foreground and background to isolate objects by converting from a gray-scale image to a binary image. Threshold can also refer to the value used to partition the image but for the sake of clarity it is fully used as threshold value.

Segmentation

It is the partitioning of an image into simpler and meaningful segments for the simplification of its analysis.

Mask

Perspective transform

Filter

hashable

hashing

Numpy Arrays

(SciPy, 2015b) In python one of the best ways to represent an array or matrix is using the numpy module. For mathematical matrices numpy provides a class called "matrix" and for arrays with element-wise operation support it provides the ndarray class. Throughout all the algorithms, arrays of the ndarray class are used extensively so it is important to understand its basic structure.

ndarray.ndim. The number of axes (dimensions or rank in python) of the array. Arrays of more than one dimension are like matrices but with a different behavior from their mathematical counterparts.

ndarray.size. The total number of elements of the array.

ndarray.shape. A tuple indicating the size of the array in each dimension. For a matrix with n rows and m columns the shape would be (n, m) therefore the length of the shape is the ndarray.ndim and the product of each element in it is ndarray.size.

ndarray.dtype. A python object describing the type of the elements in the array (i.e. all elements in an array must be of the same type). Some examples are numpy.int32 or numpy.int16 for integers, and numpy.float32 or numpy.float64 for floats, with the numbers at the end of these types being the bytes in which they are structured or can allocate data in memory but be aware that some of them differ to this naming convention (more can be read in the numpy documentation).

Numpy arrays and matrices work a little different than the operations in MATLAB but with some relations than can be easily adapted from one way to the other. These relations can be better explained in (SciPy, 2015a).

To see the common terms used addressing python code see (Python Software Foundation, 2015).

Art State

Diabetes is a chronic disease that occurs when the pancreas does not produce enough insulin or the body cannot effectively use the insulin it produces to regulate blood's sugar causing severe complications over time. These complications affects especially the nerves and blood vessels causing macro and micro vascular changes leading to increases in the risk of heart disease and stroke (50% of people with diabetes die of cardiovascular disease), renal problems, neuropathy (nerve damage) and diabetic retinopathy (DR) (Faust, Acharya U., Ng, Ng, & Suri, 2012). Type 1 diabetes (T1D, previously known as insulin-dependent, juvenile or childhood-onset) is characterized by deficient insulin production in the pancreas requiring daily administration of insulin to the patient; currently its cause is not known and it is not preventable (this does not mean that some measurements should not be taken). Type 2 diabetes (T2D, formerly called non-insulin-dependent or adult-onset) is the most chronic worldwide which comprises 90% of people with diabetes around the world (WHO, 2016; WebMD, 2014).

The rate of diabetes is increasing, according to The World Health Organization (WHO) and International Diabetes Federation (IDF) in the year 2000 the number of people with diabetes were of 171 million and it was estimated to increase to 366 million by 2030 being by then the 7th leading cause of death (WHO & IDF, 2006), but just in 2014 the number of people with diabetes rose to 422 million and caused over 1.5 million deaths worldwide in 2012. This is probably going to continue to raise as it is consistent with the global prevalence of diabetes among adults (i.e. over 18 years of age) which has risen from 4.7% in 1980 to 8.5% in 2014. This only worsens in low and middle-income countries where it is estimated that more than 80% of diabetes deaths occur and diabetes prevalence has been rising more rapidly in the last years (WHO, 2016).

Many studies concord that people with diabetes are at risk of developing diabetic retinopathy (DR) ultimately leading to blindness as a result of long-term accumulated damage to the small blood vessels in the retina. Even though people with diabetes are 25 times more likely to develop blindness and DR is considered to have caused one percent of global blindness only one-half of the patients are aware of this disease. The most effective treatment for it can only be administered at the first stages of the disease thus regular DR screening is of paramount importance and recommended annually, but most developing countries lack the ability to fully record these DR cases.

To prevent it, many models has been considered to address this problem (Askew et al., 2012). Thanks to the resent years advancements in technology and the access to it have facilitated the acquisition of advanced appliances that allow fast and reliable DR screenings in which digital imaging technology plays a key role allowing to employ state-of-the-art processing techniques to automate the detection of abnormalities in retinal images (Faust et al., 2012).

One method in particular can take huge advantages from telemedicine and digital imaging for diagnosing DR and that is the utilization of tele-ophthalmology by obtaining digital retinal images with specialized cameras (fundus cameras) and electronically transmitting them to an expert for their assessment and patient diagnosis. As a solution, it is an efficient alternative for early diagnostic and treatment of diabetic patients.

In addition, tele-ophthalmology involves the following benefits (Martínez Rubio, Moya Moya, Bellot Bernabé, & Belmonte Martínez, 2012):

- Unlike normal methods, providing healthcare to a high number of patients.
- Reducing waiting periods.
- Saving time in the diagnostic and treatment.
- Avoiding delays which can produce health problems for patients.
- Digital ocular fundus photographs are of low cost and of little discomfort for the patient.
- The application of telemedicine reduces the workload to experts letting them treat more outpatients efficiently.

Taking a particular subdivision from the ophthalmological screening methods are the non-mydriatic screening techniques. Numerous papers have demonstrated that it is one of the health interventions with the best cost-effectiveness ratio, high sensitivity and specificity. Non-mydriatic retinal cameras allow high-quality photographs through undilated pupils in digital format

offering benefits like facilitating medical compliance (i.e. patient correctly following medical advice), patients not requiring to be still at the assessment and images delivering a much broader field of view (Askew et al., 2012).

Specialized non-mydriatic fundus cameras can be cost-efficient in operational terms but exhibit a relative medium inversion cost, making them difficult to afford in some cases. Yet again, technology can offer solutions to further reduce costs while offering decent results that can be at par with what would be obtained with professional fundus cameras and that is a mobile device (e.g. cellphones, tablets, etc) which is a ubiquitous device that offers many functionalities including the acquisition of digital images and though it is not cheap it is more likely that anyone have one than a non-mydriatic fundus camera, thus preventing the necessity to acquire one.

Screening of patients with adapted mobile devices can become a highly valid method for the detection and prevention of DR allowing early access to assessments while offering easiness of use and other benefits like availability.

Programming and Research Approach

Before we discuss the algorithms, it is important to understand the context in which they were conceived. Many programming languages were taken into consideration for the implementation of the algorithms but few sufficed the requirements of high performance while keeping the readability, portability, modularity and scientific basis that a language needs. Compiled languages like C++ offer high speeds but at the cost of lengthy codes and long developing times while their contra parts, interpreted languages like MATLAB and Python, sacrifice execution speed to offer short and fast code prototyping (i.e. while developing) because of their simple yet readable languages and even offering in some cases means to improve speed by compiling or translating to other languages like C or using them as back-ends (Cython.org, n.d.; Pyzo, n.d.).

Table 1
Comparison of programming languages

Factor	C++	MATLAB®	Python
Interpreter type	compiled	just-in-time	Interpreted
Language abstraction	High-level	High-level	High-level
Standard	ISO standard	de-facto standard	de-facto standard
License	Open source	Licensed	Open source

Source: Author

Table 2
Comparison example of "for loops" in C and Python

C for loop	Python simulation of C for loop	Pythonic for loop
<pre>#include <stdio.h> #include <string.h> int main() { char s[5] = "Hello"; for(int i=0;i<strlen(s);i++){ printf("%c", s[i]); } }</pre>	<pre>s = "Hello" for i in range(len(s)): print s[i]</pre>	<pre>s = "Hello" for letter in s: print letter</pre>

Source: Author

Optimization Methods

Here some optimization techniques are explained which were used in the implementation of the algorithms.

Process big data as small data. This optimization method consist in reducing the quantity of elements in data without loosing relevant information. Some examples involve normalizing data to small values (i.e. normalize to 1) to reducing the size of arrays. In the case of images for example it can produce performance penalties if it is too big to process (e.g. 10000x8000) or not for the convenience of the algorithm. The solution is just to resize it (e.g. 500x400) using a good interpolation (though linear interpolation does fine for most cases), process the resized image and at the end resize to the original image. For data manipulation like points or domain transformations the process is similar (in fact resizing is a common transformation) and ?? ?? covers it in more detail.

Lazy evaluation. Lazy evaluations which in this case can be described as load on demand. This is advantageous to keep RAM memory free of data not in use and load them from disk when needed. In the case of structures, it is used to compute data only when it is going to be used. Python offers good solutions as the *generators* and in the case of classes the `@property` decorator to convert a method to a getter, setter or deleter useful to process a requested value on the go instead of processing everything at the class instantiation.

Preventing circular references. There is of up most importance to keep algorithms from producing object with circular references (object referenced by other objects keeping the object alive) which can produce memory leaks, instability and crashes due to exceeded memory use. In the case of python there exist a garbage collector (it can be imported as `gc`) to manage memory in the programs but currently objects with circular references cannot be collected and are kept alive until the whole program ends. To prevent it the `weakref` module can be used to provide weak references to an object which do not prevents it from being garbage collected.

Multitasking. Multitasking techniques for heavy tasks. Usually, multiprocessing was used to take advantage of multi-core computers and multithreading for lighter tasks or for task incompatible with the multiprocessing API.

Memory mapped files. It is another useful technique to keep memory free at expense of more processing time caused by memory being used not in RAM but in the local disk. This is also useful to compute big data which cannot be loaded to RAM memory and to let data be accessed for multiple tasks at the same time (Usually shared memory between programs is a really difficult achievement but it can be easily done using memory mapped files).

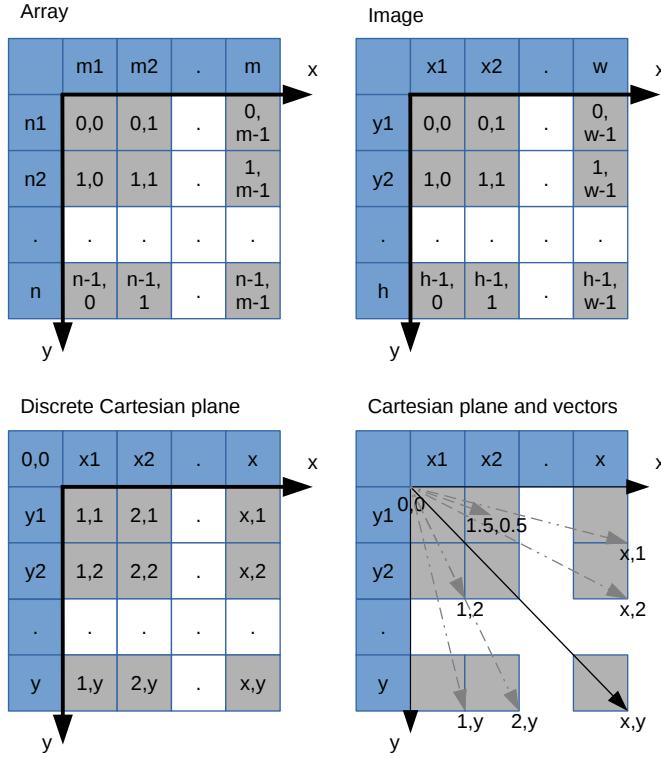
Memoization and caching. Saving data with expensive function calls or difficulty to obtain them for later use. In this way, values are cached to be used in the same session or persisted to disk to keep these computations between sessions. The persistence of data use a serialization and de-serialization process which involves saving and loading information from disk. In Python, this is achieved using the package 'pickle' and the caching using hashes to generate key-like values usually in dictionary structures (other structures can be used too).

Images in openCV

An image can be represented as a matrix of width w and height h with its elements as pixels, where each pixel is a representation of a color in one point of a discrete plane (2D dimension). In the case of OpenCV and many other libraries for image manipulation, the use of Numpy arrays as the base for image representation is becoming the standard (Numpy is a fast and powerful library for array manipulation and one of the main modules for scientific development in python). A Numpy array is multidimensional container of items of the same type and size differing mainly to the mathematical matrix in which its operations are element-wise (i.e. corresponding index in each array). The Image shape, height and width (h, w), then corresponds to a Numpy array with n rows and m columns (n, m) which in a Cartesian plane would be the (y, x) axes (Scipy, 2015).

Figure 2 shows the different representations used in this document and how they are accessed or indexed. Images and arrays are indexed from 0 to $N - 1$. Notice that images are interchangeable with arrays and can be treated like so with pixels represented by the array elements. Where the first pixel $h, w = (0, 0)$ corresponds to the first element $n, m = (0, 0)$ spanning from the origin $y, x = (0, 0)$ to the end of it where other elements being. In the context of discrete planes arrays continue to be indexed natively from 0 to $N - 1$ (the structure does not change) but are treated with an additional 1 so that arrays representing discrete planes start from 1 and finish in N . To use vectors arrays are referenced as in the discrete Cartesian case but indexes

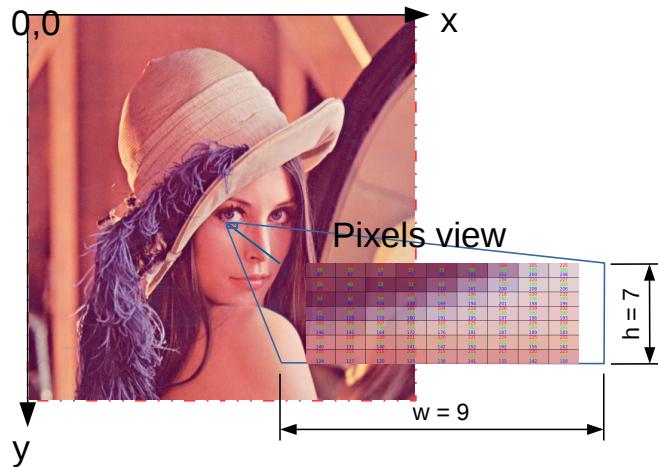
Figure 2. Used representations



are subdivided usually using interpolation so that vectors are represented using floating points but converted to integer types like in the previous cases to access the pixel's value.

Figure 3 better illustrates the notion of arrays representing images with the height as the number of rows and the width as the number of columns. Notice that from the Cartesian plane perspective the x axis goes to the right and the y axis downwards with the *origin* at the north-west of the image which corresponds to the $y, x = (0, 0)$.

Figure 3. Axes orientation of images



Algorithms design

There always emerge important questions in the development of every project like how should it be made? perhaps this piece of code is not robust enough for every possible scenario? or is it the best solution to my problem?. Well, not all solutions are absolute and there could be better implementations but for the most of it I consider that experimentation is one of the best insurance methods that exists for scientific solutions. As I see that without experimentation, a solution may not be well tested and it could not be pushed forward to the limits of their potential. Another good reason is that it opens room for creativity that could lead to unexpected inventions not yet discovered. So in the next section is presented some of the experiments that led to the implementation of the main algorithm. The repository with all the codes is in (David, 2016b). Among the repository's folders there are two containing most of the algorithms, one called *tests* where experiments were carried out and a package called *RRtoolbox* with developed algorithm used in the project. Figure 4 shows the file structure of the *RRtoolbox* package to provide a better visualization of it.

There are more algorithms provided by the *RRtoolbox* package and other non-package codes left for tasting that is not explained in next sections but the list would go so long that a book could be written about them. The other algorithms not covered here are left as task for the reader to discover in the *RRtoolbox* package or reading the manual in David2016RRtoolboxDoc. Some of them are even more complex and useful than the presented in this document but were left out for they are not relevant to the project's objectives.

General algorithms

There are many algorithms that prove useful in the manipulation of arrays that can help in the realization of more complex algorithms. These general algorithms present some of the most simple yet useful procedures for common array manipulation.

The area under a polygon (*poligonArea*). The area under a polygon can be calculated if the points of its vertices are known. This is described in (Darel Rex Finley, n.d.) which can be expressed by the general equation of the *poligonArea* as:

$$\text{poligonArea} = \frac{1}{2} |xsum - ysum|$$

Where *ysum* and *xsum* are defined as:

$$ysum = x_0y_n + \sum_{i=0}^{n-1} x_{i+1}y_i$$

$$xsum = x_ny_0 + \sum_{i=0}^{n-1} x_iy_{i+1}$$

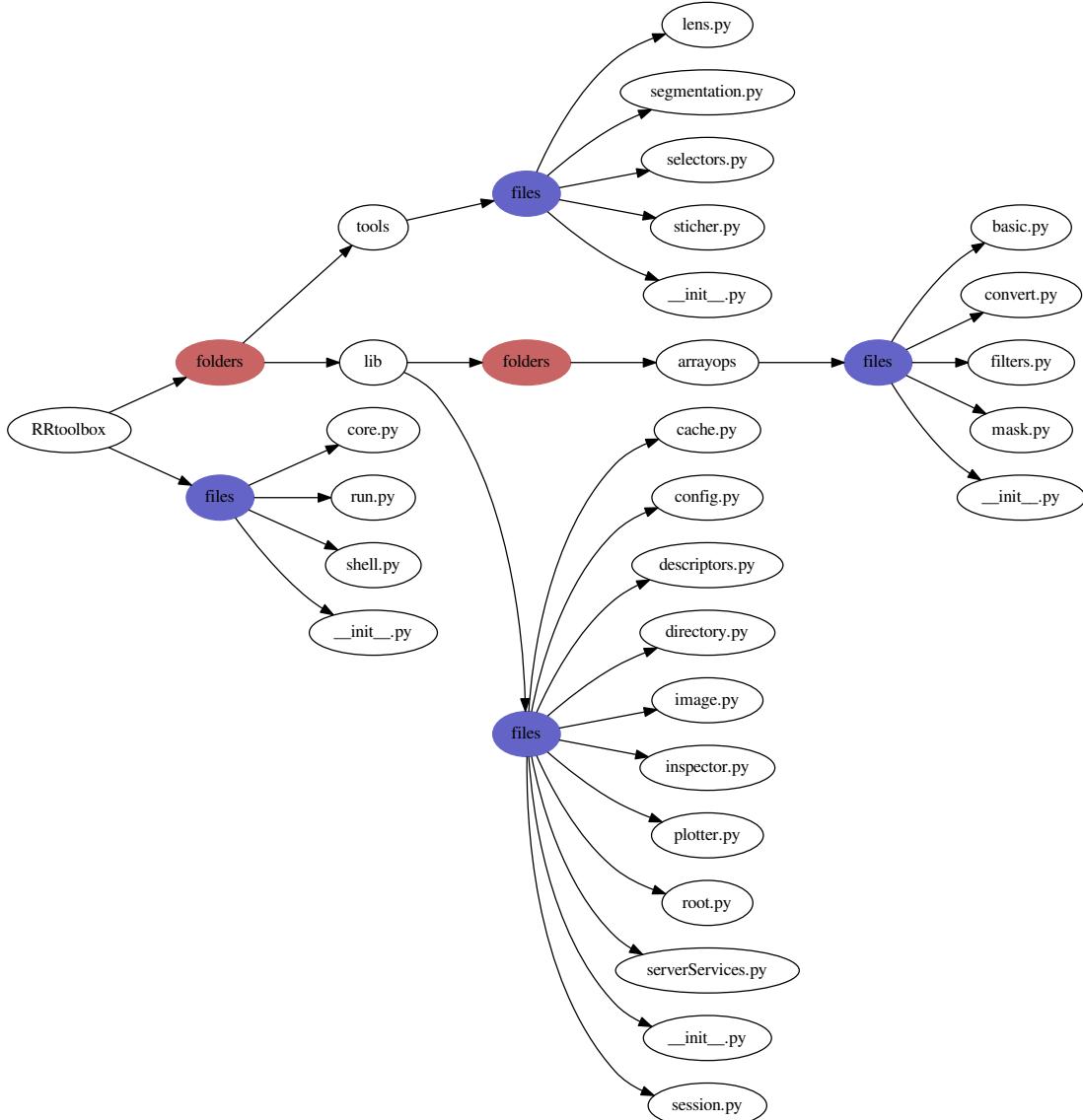
Then replacing *xsum* and *ysum* results in Equation 1:

$$\text{poligonArea} = \frac{1}{2} \left| -x_0y_n + x_ny_0 + \sum_{i=0}^{n-1} (x_iy_{i+1} - x_{i+1}y_i) \right| \quad (1)$$

This equation can calculate the area of any arbitrary polygon but with the limitation that the points that conform it must be in adequate order (i.e. if lines are traced passing at each point in order then these lines must not intersect each other). If for example we have the following points (see Figure 5):

$$\text{points} = \begin{bmatrix} -3 & -2 \\ -1 & 4 \\ 6 & 1 \end{bmatrix}$$

Figure 4. File estructure of RRtoolbox package



Then we solve it as in Equation 1 to obtain the individual summations.

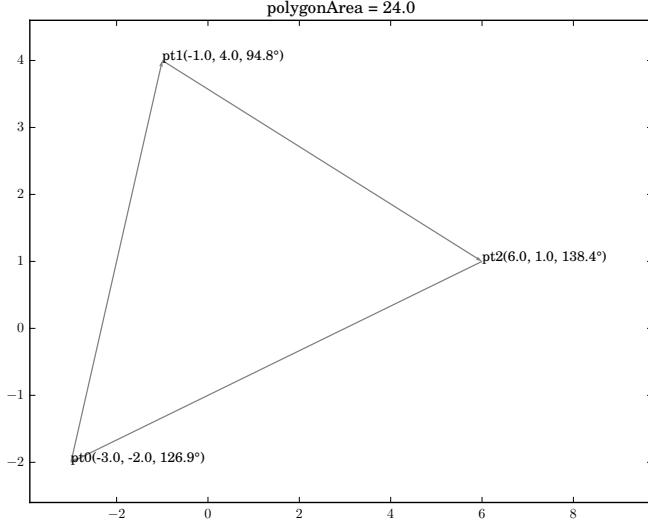
$$\begin{aligned}
 ysum &= x_0y_2 + x_1y_0 + x_2y_1 \\
 &= (-3) \cdot 1 + (-1) \cdot (-2) + 6 \cdot 4 \\
 &= 23
 \end{aligned}$$

$$\begin{aligned}
 xsum &= x_0y_1 + x_1y_2 + x_2y_0 \\
 &= (-3) \cdot 4 + (-1) \cdot 1 + 6 \cdot (-2) \\
 &= -25
 \end{aligned}$$

Which is then replaced in `poligonArea` giving the result:

$$\text{poligonArea} = \text{Abs}((-25) - 23)/2 = 24$$

Figure 5. Find the area of a polygon using Equation 1



The `polygonArea` method in Equation 1 is imported in Python as `from RRtoolbox.lib.arrayops.basic import polygonArea_calculc, polygonArea` with `polygonArea_calculc` as the actual function and `polygonArea` the alias. There are other implementations that calculates the area of a polygon and where used to provide more options and test them with each other and can be imported using `from RRtoolbox.lib.arrayops.basic import polygonArea_contour, polygonArea_fill`. The `polygonArea_contour` is a wrapper over the OpenCV method `cv2.contourArea` but with more functionalities and `polygonArea_fill` calculates the area filling the objects and summing up the number of pixels which is comparable to an invariant moment.

Overlay. This method as its name implies is used to overlay a foreground (*Fore*) over background (*Back*) array with an *alpha* transparency which can be either and array or a number. The following equation is used to with an *alpha* array:

$$\text{Overlay}_{i,j} = \text{Fore}_{i,j} \cdot \text{alpha}_{i,j} + \text{Back}_{i,j} \cdot (1 - \text{alpha}_{i,j}) \quad \forall \quad \|\text{alpha}_{i,j} \leq 1\| \quad (2)$$

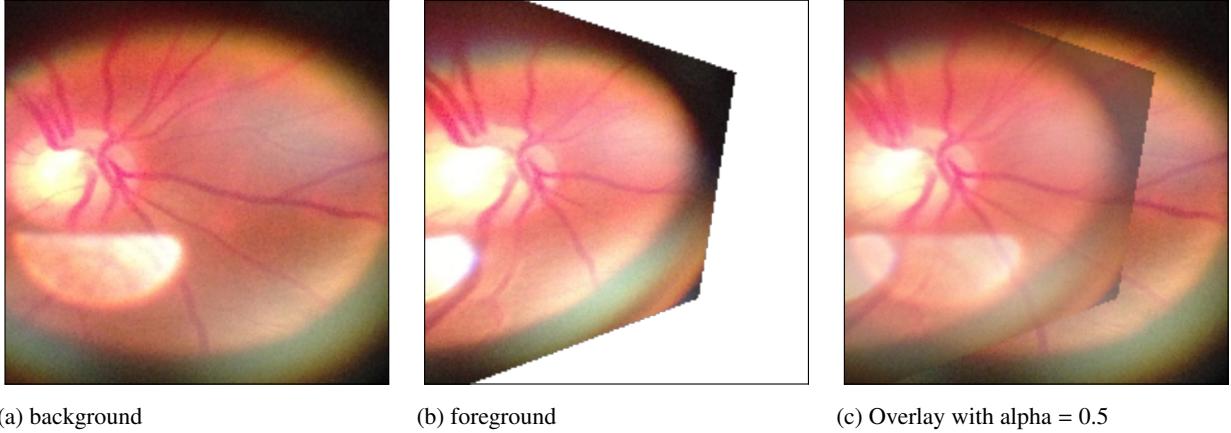
Where *alpha* is an array with numbers constrained from 0 to 1. Notice that values less than 1 attenuates *Fore* array and values more than 0 attenuates the *Back* array. This can lead to the notion of *alpha* being a transparency parameter. Equation 2 is extended using *Numpy* and is not necessarily True to operate over *Fore* and *Back* arrays of different dimensions because it uses propagation to handle arrays which uses a different mechanism complying to certain rules.

Figure 6 shows an overlay example of the implemented Equation 2 using *Numpy*. In this case *alpha* = 0.5 which is a floating number instead of an array. This function can be imported using `from RRtoolbox.lib.arrayops.basic import overlay` which offers more functionalities than in Equation 2.

Real image – Rendered image. There are situations in which we want to ‘view’ inside an image to show or select the parts of the image with more detail (Zooming-in) or that the image is too big to render in a visualization program and the solution it to resize it to a small image but conserving the coordinates of where the colors are in the original image. This algorithm was developed to comply to those needs using pure mathematics with familiar equations and some constraints.

Suppose that there exists a real image *I* with shape (H, W) for height and width respectively which has a rendered image *I_r* with shape (H_r, W_r) of any view inside *I* (i.e. any part of *I* can be selected and cropped with a different resolution than *I* that is useful for zooming-in). To calculate the coordinate system in *x, y* of the real image *I* to the coordinate system *x_r, y_r* of the rendered image *I_r* where *x_r, y_r* is a point in the domain of *I_r* and *I_r* is a ROI contained between *x₂ – x₁* and *y₂ – y₁* in image

Figure 6. Overlay example using Equation 2



I (i.e. I_r is a re-scaled ROI of I with a cropping box from the top-left point x_1, y_1 to the bottom-right point x_2, y_2) then the following equations apply:

To convert from I to I_r coordinates:

$$x_r(x), y_r(y) = W_r \cdot \frac{(x - x_1)}{(x_2 - x_1)}, H_r \cdot \frac{(y - y_1)}{(y_2 - y_1)} \quad (3)$$

To convert from I_r to I coordinates:

$$x(x_r), y(y_r) = x_1 + x_r \cdot \frac{(x_2 - x_1)}{W_r}, y_1 + y_r \cdot \frac{(y_2 - y_1)}{H_r} \quad (4)$$

With the constraints:

$$(H_r, W_r) = (y_{rmax} - y_{r0}, x_{rmax} - x_{r0}) \begin{cases} x_{r0} < x_{rmax} & \in I_r \\ y_{r0} < y_{rmax} & \in I_r \\ x_{max} \geq x_2 > x_1 \geq x_0 & \in I \\ y_{max} \geq y_2 > y_1 \geq y_0 & \in I \end{cases}$$

A complete implementation of a plotter class called *Plotim* was implemented using Equation 3 and Equation 4 for the rendering of images. This class can be used by importing it in Python with `from RRtoolbox.lib.plotter import Plotim`. Other inherited classes from *Plotim* are found in `from RRtoolbox.lib.plotter` and demonstrate the capabilities of these simple equations.

Domain transformations: scaling to original results. As explained in Real image – Rendered image section , images can be rendered differently for some needs but still processing them as the original. This is one of the techniques used to reduce processing times and standardize input images by resizing the original image I_o to a scaled image I_s which is used in the actual algorithm’s processing and then adapting the results to be applied back in the original image. But there surges a pretty obvious problem, how to convert the processed results of I_s back to I_o ? In the case of coordinates or points this can be solved using linear relations to convert from one domain to the other (explained in Real image – Rendered image section) but if the result is a transformation matrix (TM) then additional transformation matrices (TMs) must be applied to adjust it to work with the original sample.

Perspective transformation rules. Before going into mathematical detail lets define some rules and conventions useful for perspectives transformations. These conventions are used when dealing with transformations matrices:

- When the transformations are with respect to the absolute Cartesian system it is referred as xyz and all TMs must be pre-multiplied at each step.
- When the transformations are relative with respect to the previous position it is referred as uvw and all TMs are post-multiplied at each step.

The perspective transform maps from x, y coordinates to u, v coordinates using transformation matrices. To transform any x, y point in an image a 3×3 transformation matrix M is multiplied with the column vector $\begin{bmatrix} x & y & 1 \end{bmatrix}$ which yields another column vector $\begin{bmatrix} x' & y' & c \end{bmatrix}$:

$$\begin{bmatrix} M_{0,0} & M_{0,1} & M_{0,2} \\ M_{1,0} & M_{1,1} & M_{1,2} \\ M_{2,0} & M_{2,1} & M_{2,2} \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x' \\ y' \\ c \end{bmatrix}$$

Then the resulting u, v point where x, y is moved is calculated as:

$$u, v = \frac{x'}{c}, \frac{y'}{c}$$

Suppose that a scaled transformation matrix (M_s) is found to convert the perspective from a scaled image A (A_s) to the perspective of a scaled image B (B_s). So M_s is used to convert A_s to the domain of B_s but the objective is to convert the original image A (A_o) to the domain of the original image B (B_o) with a original transformation matrix (M_o). Several steps are summarized to achieve this:

1. Use a scaling transformation A_{os} to transform the points from A_o to the domain of A_s .
2. Use M_s to transform the points of A_s to the domain of B_s .
3. Use a scaling transformation B_{so} to transform the points from B_s to the domain of B_o .

Now for the mathematical abstraction.

$$\begin{aligned} M_o &= B_{so} \cdot M_s \cdot A_{os} \\ &= \begin{bmatrix} Bx_{so} & 0 & 0 \\ 0 & By_{so} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} M_{s11} & M_{s12} & M_{s13} \\ M_{s21} & M_{s22} & M_{s23} \\ M_{s31} & M_{s32} & M_{s33} \end{bmatrix} \begin{bmatrix} Ax_{os} & 0 & 0 \\ 0 & Ay_{os} & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} Ax_{os} \cdot Bx_{so} \cdot M_{s11} & Ay_{os} \cdot Bx_{so} \cdot M_{s12} & Bx_{so} \cdot M_{s13} \\ Ax_{os} \cdot By_{so} \cdot M_{s21} & Ay_{os} \cdot By_{so} \cdot M_{s22} & By_{so} \cdot M_{s23} \\ Ax_{os} \cdot M_{s31} & Ay_{os} \cdot M_{s32} & M_{s33} \end{bmatrix} \end{aligned}$$

Where

$$Ay_{os} = \frac{Ah_s}{Ah_o}, Ax_{os} = \frac{Aw_s}{Aw_o}, Bx_{so} = \frac{Bw_o}{Bw_s}, By_{so} = \frac{Bh_o}{Bh_s}$$

With (Aw_o, Ah_o) as the shape of A_o , (Aw_s, Ah_s) the shape of A_s , (Bw_o, Bh_o) the shape of B_o and (Bw_s, Bh_s) the shape of B_s . Then M_o can be rewritten as:

$$M_o = \begin{bmatrix} \left(\frac{Aw_s}{Aw_o}\right)\left(\frac{Bw_o}{Bw_s}\right)M_{s11} & \left(\frac{Ah_s}{Ah_o}\right)\left(\frac{Bw_o}{Bw_s}\right)M_{s12} & \left(\frac{Bw_o}{Bw_s}\right)M_{s13} \\ \left(\frac{Aw_s}{Aw_o}\right)\left(\frac{Bh_o}{Bh_s}\right)M_{s21} & \left(\frac{Ah_s}{Ah_o}\right)\left(\frac{Bh_o}{Bh_s}\right)M_{s22} & \left(\frac{Bh_o}{Bh_s}\right)M_{s23} \\ \left(\frac{Aw_s}{Aw_o}\right)M_{s31} & \left(\frac{Ah_s}{Ah_o}\right)M_{s32} & M_{s33} \end{bmatrix} \quad (5)$$

A family of applications and functions are found in the *RRtoolbox* package using the concept of converting data for convenience and the reconverting results to the intended target. In particular many of these methods can be found in *RRtoolbox.lib.arrayops.convert* where the particular case presented in Equation 5 can be imported in Python as `from RRtoolbox.lib.arrayops.convert import sh2oh`.

Normalization. Array normalization is useful to convert all the values to a more comfortable representation or to standardize them. An array A is normalized to the range $[0, 1]$ with the simple function:

$$N(A) = \frac{A - \min(A)}{\max(A)} \quad (6)$$

Where \min and \max are functions that return the minimum and maximum element of A respectively. The normalization is imported in Python using `from RRtoolbox.lib.arrayops.basic import normalize, normalize2, rescale` with `normalize2` and `rescale` being variants of the normalization.

Load functions

In python there are several ways to load an image with functions from packages that provide different features making it easy to provide support to load images from different sources, formats and specifications. This capability was exploited by coding a way to seamlessly load images when provided the path to the source and determine the means to retrieve them. A functional implementation can be imported using `from RRtoolbox.lib.image import loadFunc` which creates a loader function to use when quickly loading many images with the same loading configuration or to keep images in disk and load them on the fly with the desired configuration (i.e. with certain shape, any supported color transformation and from desired sources like URLs, sockets, memmapped files or supported formats). The `loadFunc` in most of the implementations of the *RRtoolbox* package and has become kind of a standard in it. More useful functions and classes are found in *RRtoolbox.lib.image*.

Load from sockets. Loading an image from a socket is achieved using the `socket` package from python by defining a server and a client with a custom simple protocol to transfer pickled data.

Load from URLs. To load images from an Uniform Resource Locator the `urllib` family of modules from python can be used. In python there is `urllib`, `urllib2` and `urllib3` with some differences between python 2 and python 3. For the implementation it was used `urlopen` from `urllib3` and `urllib.request` from python 2 and 3 respectably which returns a handle similar to a file-like object created by the built-in `open` object used to open image's URLs. These two APIs being similar facilitates more the implementation.

Load from files. It can be achieved directly using the OpenCV function `cv2.imread` or loading images' files using the build-in `open` function, retrieving the binary string, converting it to an array with `np.fromstring` and then decoding the resulting array with `cv2.imdecode` to obtain the image.

Load from memmapped files. This is one of the ways to save memory to a file and access that file directly from the disk instead of keeping it in the RAM. To create and retrieve a binary file in *NumPy* format (i.e. extension `.npy`) the save and load functions are provided in the `numpy.lib` module which creates a binary representation of a *NumPy* array that can be simply loaded to memory, pickled (technically it is the pickled file itself different to the string representation used for other purposes e.g. store in databases as a string and not as a external file), transferred remotely, accessed with standard read and write modes and still behave like a normal *NumPy* object which can be referenced or shared between processes (useful for multitasking).

Pre-processing, Filters and Enhancing Methods

Histogram equalization. Histogram equalization (*HE*) distributes the colors equally in an image based from its histogram, so if an image is bright most pixels will be confined to the highest levels or if an image is dark most pixels will be at the lowest levels and *HE* will distribute this levels to the levels that are lacking pixels improving the contrast of the image. That way images of the same scenery with different light conditions will be almost the same after image equalization or histograms confined to particular regions will be distributed to all regions (K & Mordvintsev, 2013).

To cover the full spectrum a transformation function or lookup table is needed to map from the input pixel of the specific regions to the output pixels of the full region. This is accomplished using the cumulative distribution function (*CDF*), in a sense *HE* helps standardize the *CDF* of an image when distributing all the colors equally throughout it. Code 1 is a MWE which demonstrates the histogram equalization. Additional explanations in (Gonzalez, Woods, & Eddins, 2004, section 3.3.2).

Code 1: Histogram equalization MWE

```
import cv2
import numpy as np
img = cv2.imread('image.jpg',0) # load the image
hist,levels = np.histogram(img.flatten(),256,[0,256]) # get histogram
cdf = hist.cumsum() # get CDF by accumulating histogram
cdf_m = np.ma.masked_equal(cdf,0) # mask to convert to range [0,1]
cdf_m = (cdf_m - cdf_m.min())*255/(cdf_m.max()-cdf_m.min()) # scale to range [0,255]
cdf = np.ma.filled(cdf_m,0).astype('uint8') # mask as a look-up table
img2 = cdf[img] # apply the mask to transform the inputs pixel to output pixels
```

OpenCV has a function to apply *HE* called `cv2.equalizeHist` where its input is a grayscale image and the output is the equalized image.

A drawback is that it won't work well where there are large intensity variations, if for example a histogram is bimodal with the two peaks apart of each other (i.e. one peak at low levels and the other at high levels) or histogram is not confined to a particular region then these will be equalized causing lost of details in some cases due to saturation (where there is saturation lost of information is present too). To prevent it openCV provides a Contrast Limited Adaptive Histogram Equalization (*CLAHE*) function that is applied to sub-regions in the image. This approach does not consider the global contrast of the image as in *HE* preventing the problems that this implies, in contrast *CLAHE* divides the image into small blocks called *tiles* of typically 8x8 (the parameter is `tileGridSize`) to apply *HE* in each one and that way the histogram would be confined to small regions (solving the particular problems in *HE*) but this brings up another problem, if noise is present it will be amplified. To avoid this, a contrast limit is applied to any level above a specified contrast (the parameter is `clipLimit` with 40 as its default value) in a process where the pixels are clipped and distributed uniformly to other levels before *HE* is applied. Bilinear interpolation is applied to remove artefacts created at the borders of each tile after each individual equalization (K & Mordvintsev, 2013).

In general *HE* techniques are used in pre-processing operations to usually normalize the image patterns and lighting conditions (Médioni, 2005) but it has been observed that it intensifies noise (Antal & Hajdu, 2012) and for that noise attenuation is recommended before using *HE* (Sopharak, Uyyanonvara, & Barman, 2013). For further optimizations using C++ and CUDA to process in the computer's GPU you can follow the advices in (Fierval, 2015).

Figure 7 shows a *HE* and *CLAHE* comparison with their respective histogram of colors. Notice the input image (Figure 7a) and how the *HE* equalizes the input (Figure 7c) by converting its non-linear *CDF* (Figure 7b) in an uniform *CDF* (Figure 7d). This leads to some uneven parts in the image but *CLAHE* solves it as explained before (Figure 7e).

Figure 8 shows additional experiments with the *CLAHE* method using color images. The default application with parameters `tileGridSize` = (8, 8) and `clipLimit` = 2.0 is found in Figure 8b. In Figure 8c the parameters `tileGridSize` = (30, 30) and `clipLimit` = 2.0 are used and in Figure 8d changed to `tileGridSize` = (8, 8) and `clipLimit` = 15.0. This example demonstrates that in color images the application of *CLAHE* with an incorrect selection of parameters can lead to unappealing results.

Matrix decomposition to reduce levels. Matrix decomposition offers a different representation of the original image array and lets the information to be changed easily while affecting to all pixels instead to a single one. This can reduce the diversity of image levels to offer less colors than the original and reduce file storage but at an unwanted price, less details and added artefacts like edges that can prevent the image to be adequately computed. Code 2 shows an implementation using *NumPy* and the developed package *RRtoolbox* explained in other sections.

Figure 7. Histogram equalization example

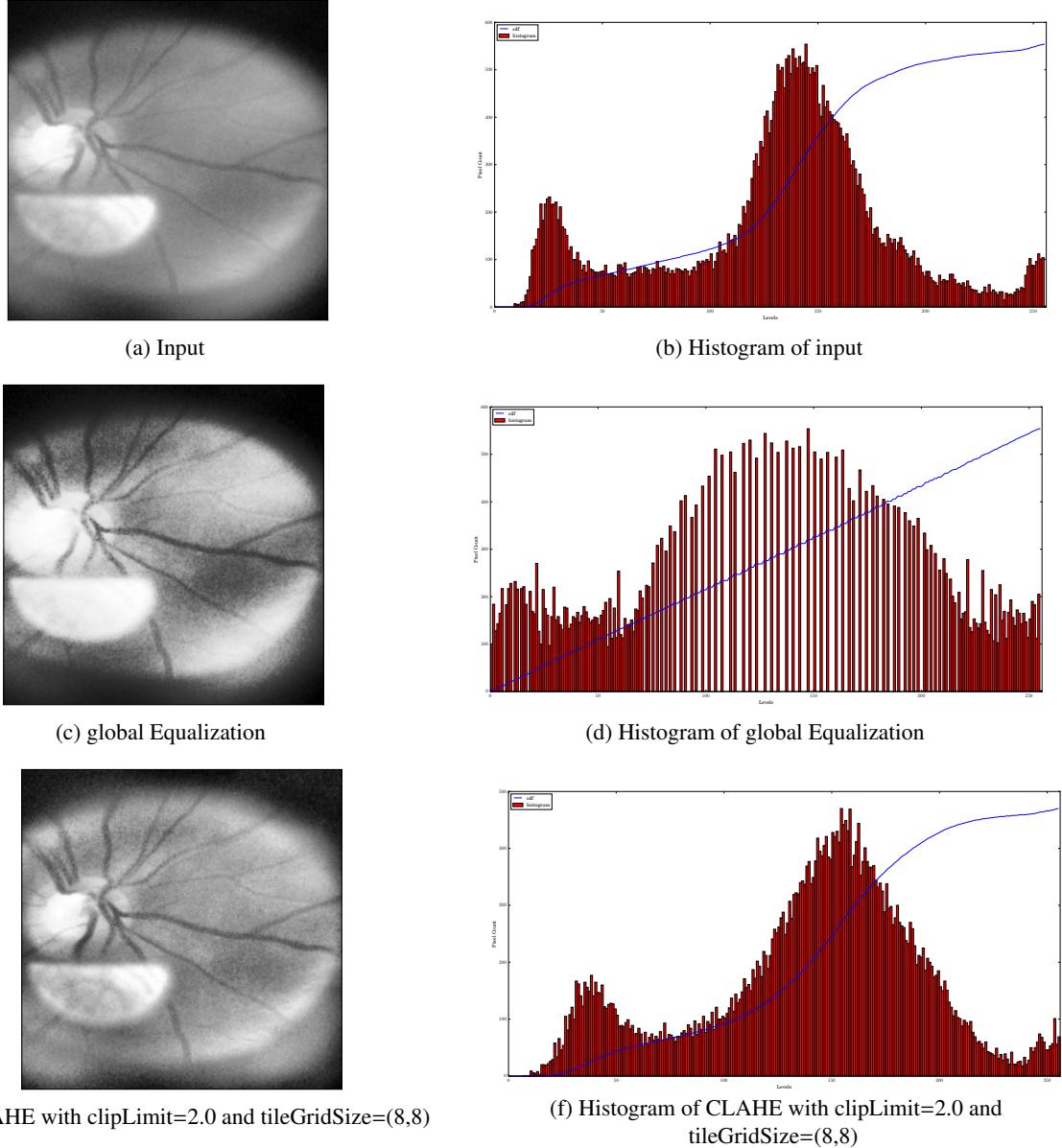
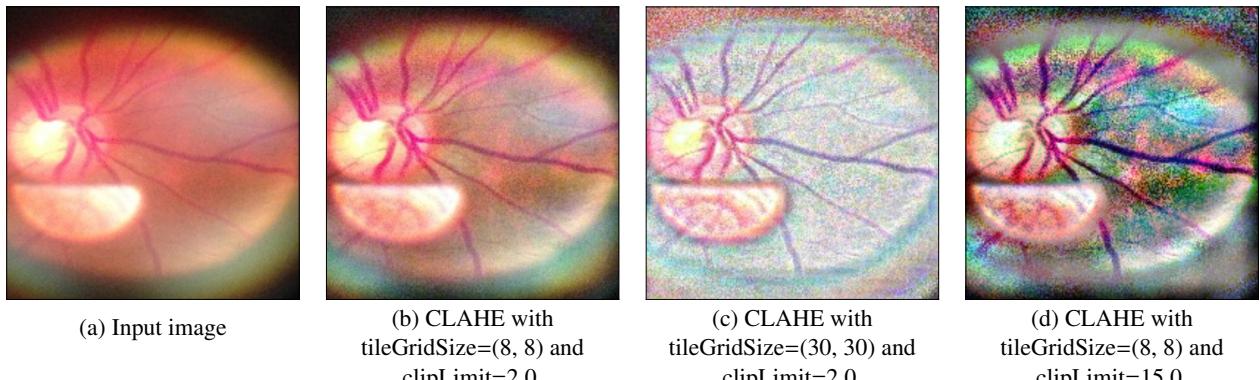


Figure 9 shows two decomposition tests for a gray image with different patterns and intensities (Figure 9a). Notice that the shape of the s array holding the singular values is $s.shape = (200,)$ corresponding to the minimum value of the image's shape $im.shape = (200, 200)$. The recomposed image with slice $s[50:] = 0$ produces an almost complete original image (Figure 9b) while with slice $s[1:] = 0$ produces an background-like image without patterns (Figure 9c) and with slice $s[:1] = 0$ only the patterns or edges of the original image are preserved (Figure 9d).

Smoothing with 1D-filters. There is a filter algorithm in (SciPy, n.d.) presenting the 'flat', 'hanning', 'hamming', 'bartlett' and 'blackman' windows that are used to convolve with the input data. This operation smooths out data which exhibit noisy like behaviour but with trends to certain values. One of their applications is to smooth histograms to obtain their local minima and maxima values.

As a reference filter Figure 10 shows the Savgol or Savitzky-Golay filter which uses a method much like a fitting algorithm combined with a convolution. This filter is used to compare the other filters due that it fits almost perfectly to the histogram

Figure 8. CLAHE



Code 2: Decomposition MWE

```

from RRtoolbox.lib.image import loadcv, np
W,H = 400,300 # This is how shape works with Numpy H,W == im.shape = rows,cols
shape = W,H # this is how functions receive shape

# load the image
im = loadcv("my_image.jpg",0,shape)

# get the Singular Value Decomposition (SVD) of the image
U, s, V = np.linalg.svd(im, full_matrices=False)
# U - Unitary matrices. The actual shape depends on the value of
# s - The singular values for every matrix, sorted in descending order.
# V - Unitary matrices

# reduce information in decomposition
print s.shape # this correspond to np.min(shape) and s rows go from 0 to s.shape-1

# recompose original image
im = np.abs(np.dot(U, np.dot(np.diag(s), V))) # complete SVD

# copy s array for two tests
s1 = s.copy()
s2 = s.copy()

s1[:1]=0 # destroy the values at the most significant row 0

# recompose image 1
im1 = np.abs(np.dot(U, np.dot(np.diag(s1), V))) # reduced SVD

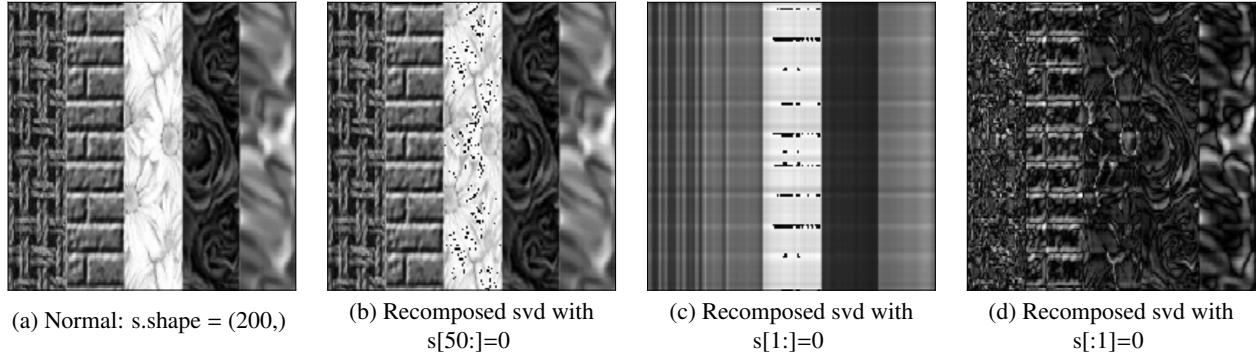
s2[50:]=0 # destroy the values of the less significant
# rows from rows 50 to the last row s.shape-1

# recompose image 2
im2 = np.abs(np.dot(U, np.dot(np.diag(s2), V))) # reduced SVD

```

sample. Notice that this prevents the filtered data to not be smoothed as desired to reduce the number of local minima and maxima values and the incorrect parametrization of the filter can lead to reduce the fidelity of the filtered data (Figure 10b).

Figure 9. Decomposition tests



This filter can be found in the *Scipy* Python package using `from scipy.signal import savgol_filter`. $window_{len}$ is the length of the filter window (i.e. the number of coefficients in the polynomial) and must be a positive odd integer. $polyorder$ is the order of the polynomial used to fit the samples and must be less than $window_{len}$.

Figure 10. Savgol filter

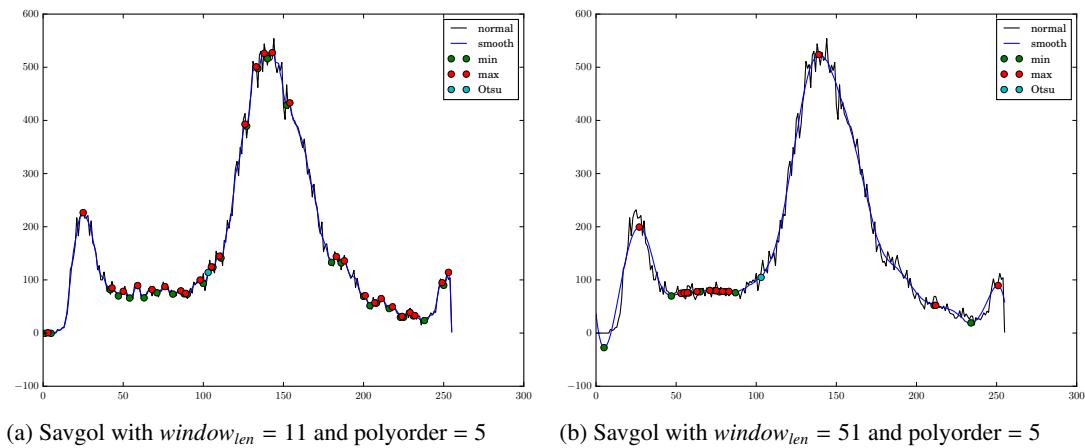


Figure 11 shows the filter as implemented in (SciPy, n.d.) using the Hanning window. Filtered data is moved proportional to $window_{len}$ creating an offset between the input and the filtered result. This produce a wrong result for the minima and maxima values obtained using the filtered data which should be aligned with the input data.

Figure 12 shows the improved filter using the Hanning window with shifted convolution to correct filtered data offset presented in Figure 11. Notice that filtered data does not present offset between it and the input and the minima and maxima values correspond to the actual values of the input. The proposed filter is found in the developed *RRtoolbox* package and it can be imported using `from RRtoolbox.lib.arrayops.filters import smooth`.

Gaussian Filter. It is one of the most widely used filters to reduce Gaussian noise in images but it produce a blur-like effect that eliminates details along with its application. Nonetheless, it is known that this can be a desirable effect used to simulate the loss of detail produced when an scenery is photographed from a farther viewpoint for applications where scale invariance is important (Rey Otero & Delbracio, 2014).

Bilateral Filter. The *bilateralFilter* can reduce unwanted noise very well while keeping edges fairly sharp. However, it is very slow compared to most filters so it is not recommended for real-time applications where the hardware is limited and can't cope with fast processing times giving retarded results. In the other hand, this filter does very well in applications where the

Figure 11. Filter with Hanning window

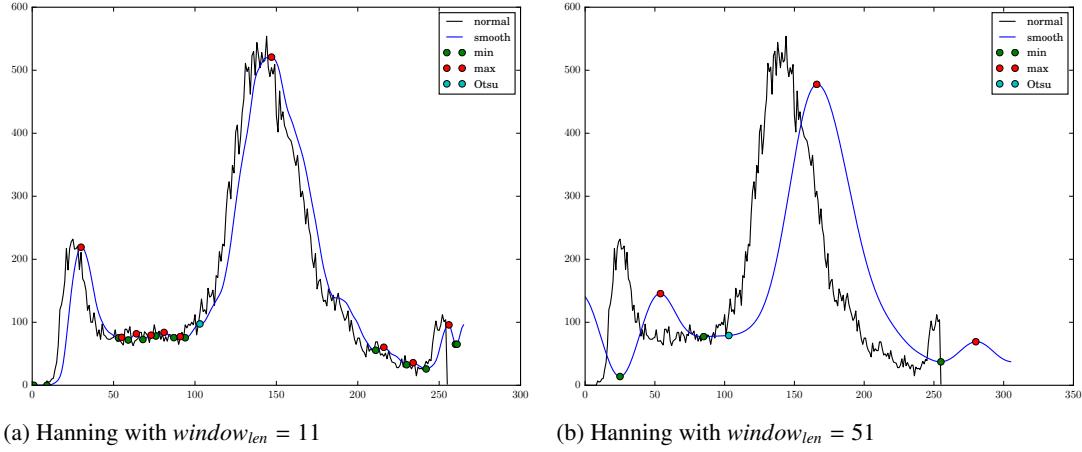
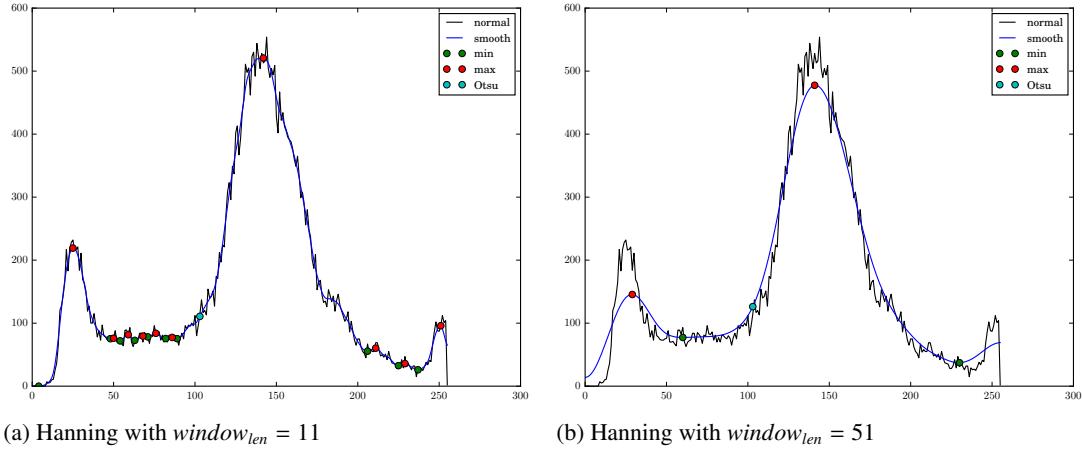


Figure 12. Filter with Hanning window and shifted convolution



time is not the most important factor for its functionality. The OpenCV package offers the function *bilateralFilter* (OpenCV, n.d.-b) which applies bilateral filtering to the input image as described in (Paris, Kornprobst, Tumblin, & Durand, 2008) with the following arguments:

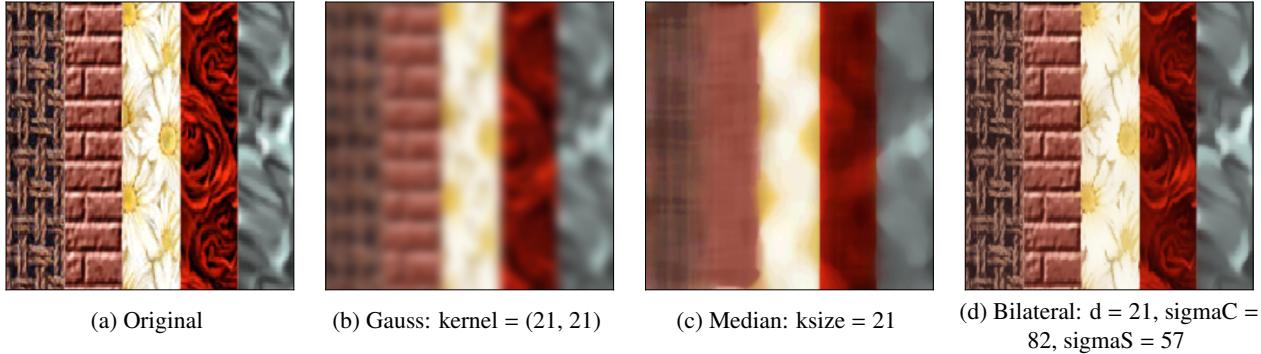
- *src*— Source 8-bit or floating-point, 1-channel or 3-channel image.
- *dst*— Destination image of the same size and type as *src*.
- *d*— Diameter of each pixel neighbourhood that is used during filtering. If it is non-positive, it is computed from *sigmaSpace*.
- *sigmaColor*— Filter sigma in the color space. A larger value of the parameter means that farther colors within the pixel neighbourhood (see *sigmaSpace*) will be mixed together, resulting in larger areas of semi-equal color.
- *sigmaSpace*— Filter sigma in the coordinate space. A larger value of the parameter means that farther pixels will influence each other as long as their colors are close enough (see *sigmaColor*). When *d* > 0, it specifies the neighbourhood size regardless of *sigmaSpace*. Otherwise, *d* is proportional to *sigmaSpace*.

Filter size. Large filters ($d>5$) are very slow, so it is recommended to use $d=5$ for real-time applications, and perhaps $d=9$ for off-line applications that need heavy noise filtering.

Sigma values. For simplicity, you can set the 2 sigma values to be the same. If they are small (<10), the filter will not have much effect, whereas if they are large (>150), they will have a very strong effect, making the image look "cartoonish".

Figure 13 shows a comparison between the bilateral filter and the most common spacial filters, the Gaussian and Mean filter. As it is known the Gaussian filter (Figure 13b) produces a blurry-like result from the original (Figure 13a) and the Mean filter eliminates most of the edges from it (Figure 13c) while the bilateral filter preserves the edges and retains most of the features (Figure 13d)

Figure 13. Spacial filters comparison



The bilateral filter tries to preserve the edges but not always retains the features if it is not well parametrized which could be an unwanted result for the user. To prevent this some tests were made using the filter and getting good parameters to make a selector function for the correct bilateral filter parameters based in the image shape and the 'mild', 'normal' and 'heavy' modes of noise in the image. Figure 14 shows the selection for the parameters in the Bilateral filter according to the image's shape and mode. Notice that the mode only changes the diameter parameter d and the others continue to variate equally when the shape of the image changes too. The selector function is found in the developed *RRtoolbox* package imported as `from RRtoolbox.lib.arrayops.filters import getBilateralParameters`.

Figure 15 shows the results of the Bilateral filter for an noisy image with different shapes with the parameters selected for the algorithm as shown in Figure 14 in 'heavy' mode. Notice here that when images are bigger and bilateral filters compensate for it to filter noise it consume a considerate amount of time to filter.

SigmoidImageFilter. The sigmoid filter is based in the sigmoid function which is characterized by having an "S" shape with a dampening behavior when tending to $-\infty$ and an approximation to the value "1" when tending to ∞ (i.e. an adequate offset can be used to determine the threshold to damper or let through certain values). In particular this filter was inspired from one of the tools of Mevislab, the *SigmoidImageFilter* which uses the Insight Segmentation and Registration Toolkit (ITK) described in (ITK, n.d.). The equation is as follows:

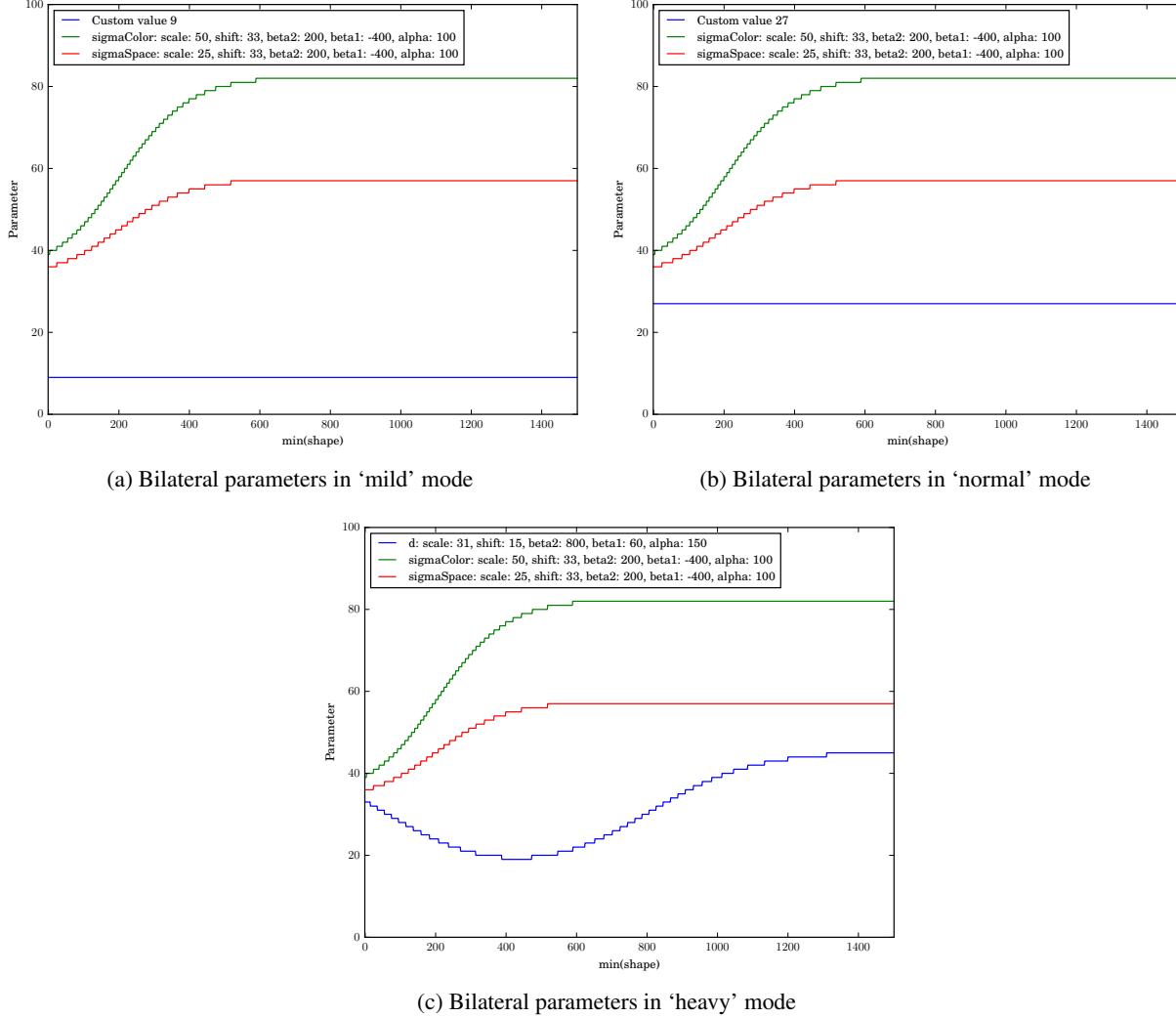
$$\text{Sigmoid}(x_i, \alpha, \beta, \text{Max}, \text{Min}) = (\text{Max} - \text{Min}) \left(\frac{1}{1 + e^{-(x_i - \beta)/\alpha}} \right) + \text{Min} \quad (7)$$

where Max is the maximum value at the output, Min is the minimum value at the output, β the offset and α a scaling factor. The β parameter can be thought as the offset on the pixel value that you are trying to isolate i.e. if the object you are trying to segment is at a pixel intensity above 150, you would choose a β value that is around that value and the α parameter can be thought of as the scaling or variance of the sigmoid (CMISS, n.d.).

Normalized SigmoidImageFilter. The normalized sigmoid filter is extracted from the component in Equation 7 that computes any x value (for discrete it is x_i) with an offset β and a scaling factor α that its result is inside the range of $[0, 1]$. This is:

$$nS_i = \text{normSigmoid}(x_i, \alpha, \beta) = \frac{1}{1 + e^{(\beta - x_i)/\alpha}} \quad \forall x_i, \alpha, \beta \in \mathbb{R} \quad \wedge \quad 0 \leq nS_i \leq 1 \quad (8)$$

Figure 14. Selection of bilateral parameters according to image shape



which stabilizes at the limits:

$$\begin{aligned} \lim_{x \rightarrow -\infty} nS(x) &= 0 \\ \lim_{x \rightarrow \infty} nS(x) &= 1 \quad \forall \alpha > 0 \end{aligned}$$

This facilitates the calculation of α and β . For example, solving for α yields:

$$\alpha(\beta) = \frac{\beta - x}{\ln\left(\frac{1}{nS} - 1\right)}$$

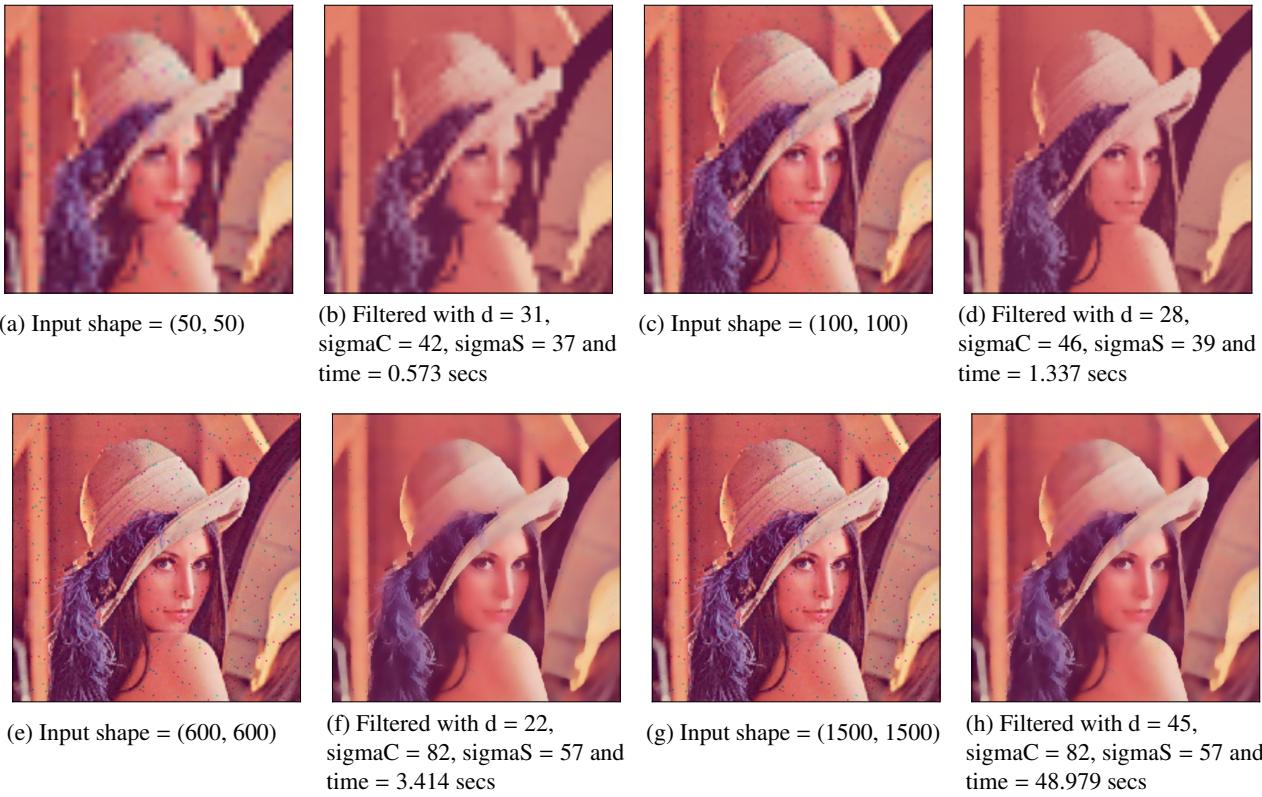
And solving for β yields:

$$\beta(\alpha) = \alpha * \ln\left(\frac{1}{nS} - 1\right) + x$$

Also, Equation 7 can be expressed in terms of the Equation 8 as follows:

$$\text{Sigmoid}(x, \alpha, \beta, \text{Max}, \text{Min}) = (\text{Max} - \text{Min}) * \text{normSigmoid}(x, \alpha, \beta) + \text{Min} \quad (9)$$

Figure 15. Bilateral filter exposed to different shapes



which stabilizes at the limits:

$$\lim_{x \rightarrow -\infty} Sigmoid(x) = Min \quad \forall \alpha > 0$$

$$\lim_{x \rightarrow \infty} Sigmoid(x) = Max$$

The sigmoid function and Normalized sigmoid are in the *RRtoolbox* package and can be imported using `from RRtoolbox.lib.arrayops.filters import sigmoid, normsigmoid`.

Custom filters using *normSigmoid*. Observing that the *normSigmoid* function is similar to a Butterworth filter it can be used to make custom filters as in signal processing allowing an image to be filtered (as in non-spacial or color filters) with common filters such as *high-pass*, *low-pass*, *band-stop*, *band-pass* or any custom filter created by the combination of the *normSigmoid* function and normalizations (see Equation 6).

The simplest filters can be made using a single *normSigmoid* function. These are the high-pass and low-pass filters created when α is positive and negative respectively. The others can be seen as a compound of these two.

$$Lowpass(x, \alpha, \beta) = normSigmoid(x, \alpha, \beta) \quad \forall \alpha < 0 \quad (10)$$

$$Highpass(x, \alpha, \beta) = normSigmoid(x, \alpha, \beta) \quad \forall \alpha > 0 \quad (11)$$

$$Bandstop(x, \alpha, \beta) = Lowpass(x, \alpha, \beta_1) - Lowpass(x, \alpha, \beta_2) + 1 \quad \forall \beta_1 > \beta_2 \quad (12)$$

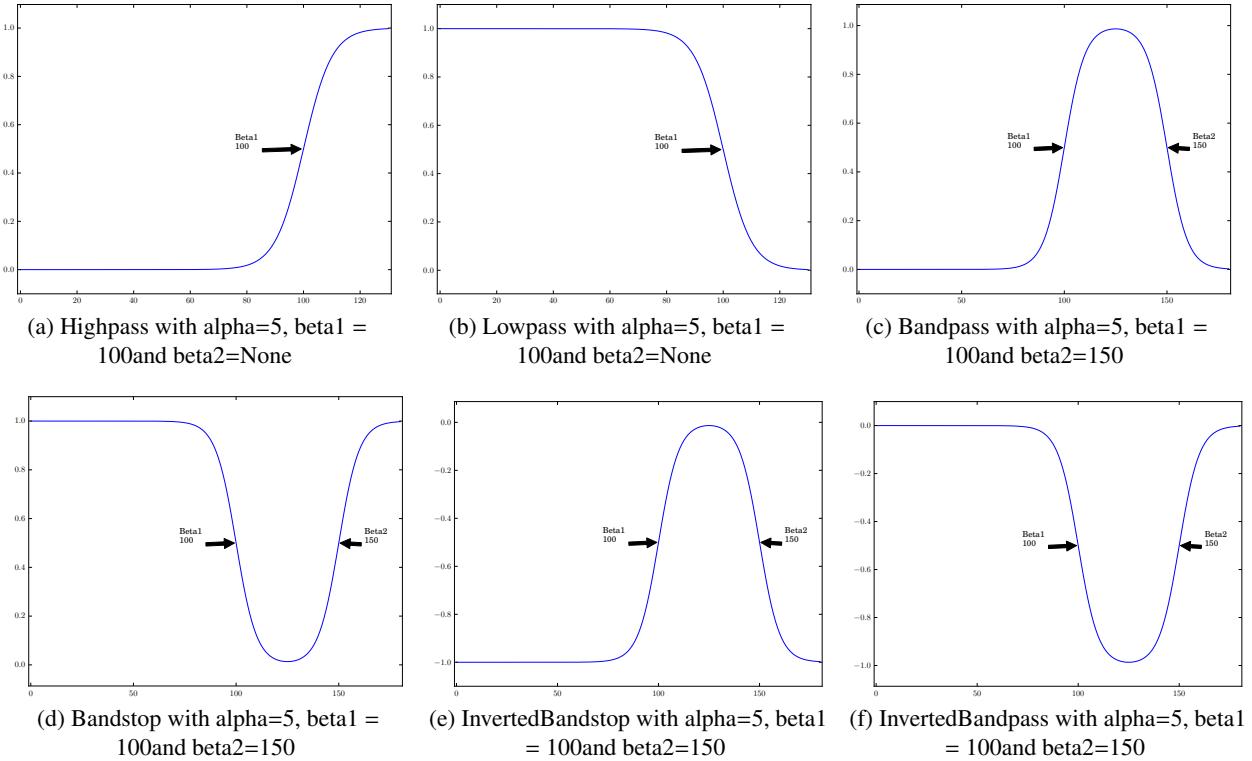
$$\text{Bandpass}(x, \alpha, \beta) = \text{Highpass}(x, \alpha, \beta_1) - \text{Highpass}(x, \alpha, \beta_2) \quad \forall \quad \beta_1 > \beta_2 \quad (13)$$

$$\text{Bandstop}_{\text{Inverted}}(x, \alpha, \beta) = \text{Lowpass}(x, \alpha, \beta_2) - \text{Lowpass}(x, \alpha, \beta_1) - 1 \quad \forall \quad \beta_1 > \beta_2 \quad (14)$$

$$\text{Bandpass}_{\text{Inverted}}(x, \alpha, \beta) = \text{Highpass}(x, \alpha, \beta_2) - \text{Highpass}(x, \alpha, \beta_1) \quad \forall \quad \beta_1 > \beta_2 \quad (15)$$

Figure 16 shows the response of each normalized filter and Figure 17 the comparison among them. Notice how all the filters converge if they share the same α and β parameters. These filters are imported using `from RRtoolbox.lib.arrayops import FilterBase, Lowpass, Highpass, Bandstop, Bandpass, InvertedBandstop, InvertedBandpass, filterFactory`. The additional import, `ffilterFactory`, is a function used to create the necessary filter from the provided arguments.

Figure 16. Common filters response using *normSigmoid*



All the filters which range are between [0, 1] are called normalized filters and can be derived from *normSigmoid* or the *high-pass* and *low-pass* filters. This inheritance-like behaviour led to the normalized filters being implemented using classes. The class diagrams can be seen in Figure 18. It shows how the filter classes inherit from the *FilterBase* class, their methods (represented by the `m` icon) and fields (`o` icon). The `c` icon is for classes, `x` deleter, `g` getter and `s` setter properties. As the normalized filters suggest the color levels in the image I are saturated when passed through them:

$$\text{filtered}I_{\text{saturated}} = \text{filter}_{\text{normalized}}(I)$$

To filter the image I without saturation (only dampened values are saturated to zero) a simple element-wise matrix multiplication can be carried out as follows:

$$\text{filtered}I_{\text{un-saturated}} = \text{mul}(I, \text{filter}_{\text{normalized}}(I)) = I_{i,j} * \text{filter}_{\text{normalized}}(I_{i,j})$$

Figure 17. Common filters comparison

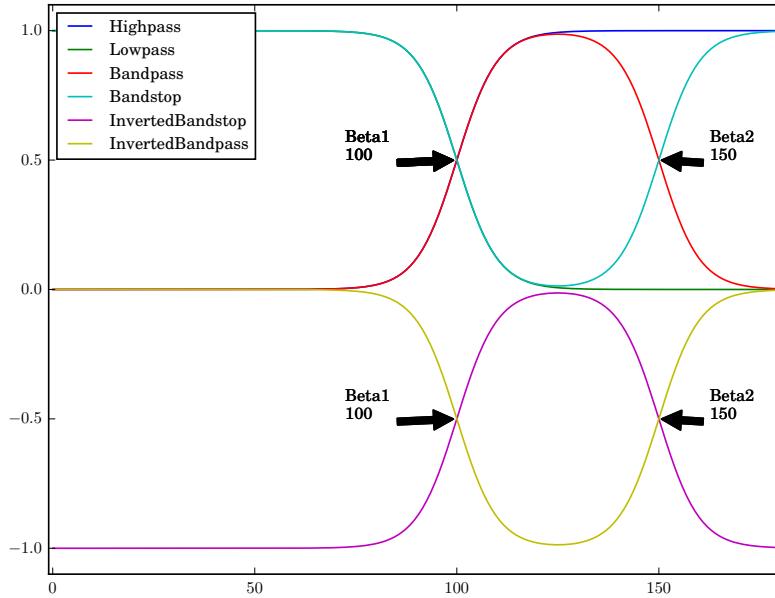
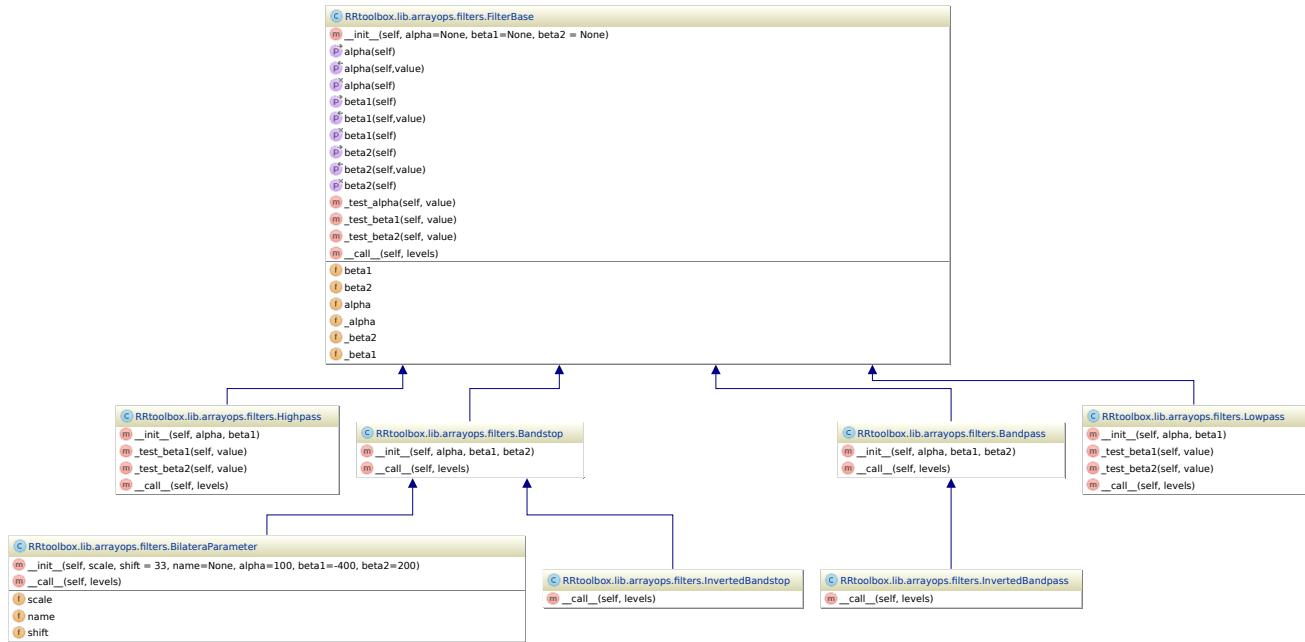


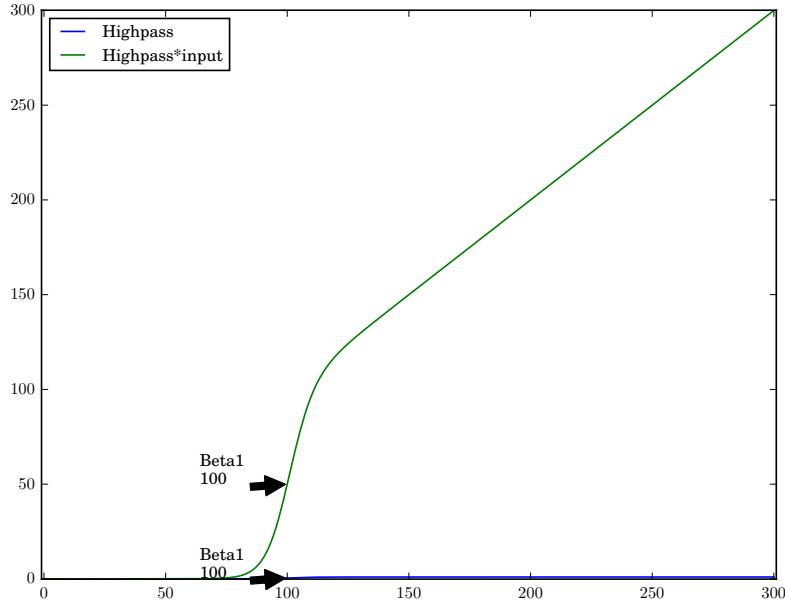
Figure 18. Filters class diagrams



Where the `mul()` function denotes element-wise multiplication or index by index operation which is different to matrix multiplication in mathematics.

Figure 19 shows an example of the normalized *high-pass* filter customized to prevent saturation. Filters that are not comprised between 0 and 1 are referred as custom filters. Notices that the response tends to infinite when input tends to infinite too. Because images of *uint8* type values have 256 levels ranging from 0 to 255 the filter only would filter those values and would not have problems with the unstable behaviour.

Figure 19. Custom filtering without saturation



Sigmoid filtering and saturation.

Segmentations

Several segmenting methods were taken into account but only threshold methods were developed for simplicity. The principal idea in segmenting retinal images is to over-thresh and further process them according to some general observations in the experimentation to obtain the final segmentation. Some methods were compared against an expert (i.e. what is considered to be a right threshold) and others to a well known automated threshold method, in this case the *Otsu* method. This with the intention of examining the algorithm robustness and reliability when applying to an unknown retinal image. Next subsections assume that a general threshold is applied to an image and an specific problem wants to be solved.

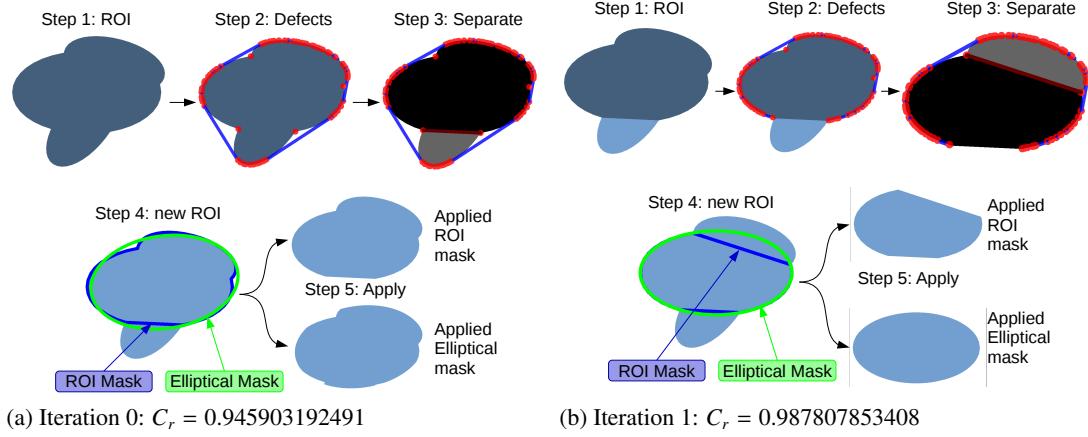
Convex hull with line cuts. This method separates an irregular object where two convexity defects are dominant, that is were the distances between the object and the convex hull are more pronounced. This is intended to be applied on objects with irregularities or protuberances where what is wanted are regular shapes like squares, circles and ellipses or objects with few defects.

The iterative splitting of the object can be stopped using the convexity ratio $r_{convexity}$ explained in the Convexity ratio section .

The pseudo code is as shown in ???. Its implementation is not part of the *RRtoolbox* package as it is still a concept not adequate for robust operations. It can be found under *tests/hypothesis2_5_masks_defect_lines2.py* in the repository.

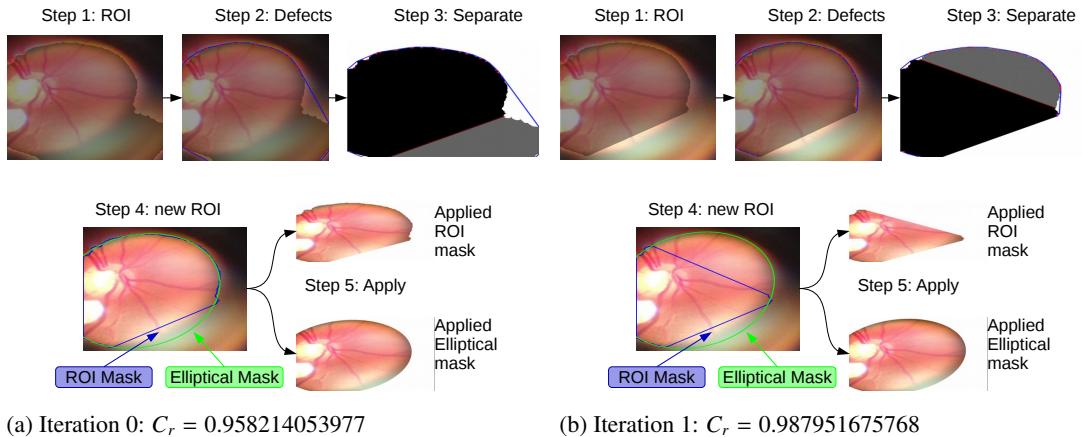
This concept in *tests / hypothesis2_5_masks_defect_lines2.py* is demonstrated using a model object as in Figure 20. The region of interest (ROI) is an irregular object with an obvious round main body and some humps sticking from it (it 1, step 1). These humps form defects with respect to the convex hull of the ROI (it 1, step 2) so the more pronounced the defect is the greater the distance from the ROI and the convex hull is. This way it can be detected where the biggest irregularity is by selecting the two most biggest distances from the ROI and the convex hull and finding their points in the binary image. After this, the main body can be easily segmented by cutting the ROI in two by these two points (it 1, step 3). When the ROI is divided in two it is the choice of user to select the new ROI but for the sake of automation here and because what is wanted is easy enough as to know which object is bigger then a simple comparison is made to select the main body. As expected the main body is the object with biggest area (it 1, step3, colored in black) and the other (in gray) is a protuberance which is discarded. Once the new ROI is selected it can be used directly as a mask to apply it in the object (it 1, step 5, above), refined it to better match the object (it 1, step 5, below) or it can be further developed by applying the same algorithm in another

Figure 20. Ideal threshold with defect lines



iteration and sub-segment it until the desired ROI is reached (i.e. one way to stop the iterations is by using a goal with $r_{convexity}$ which approach to 1 the more it approach to its convex hull). Using the new ROI and iterating over it a second time though the same steps eliminates the second hump in the original ROI (it 2, step 4, ROI mask) producing an accumulative effect each time a new iteration is made. Likewise, as stated, the $r_{convexity}$ which was 0.945 in the iteration 1 and now is 0.987 in the iteration 2 is approaching to a ratio of 1 giving a good indication that the object has been successfully approached to its convex hull or a more regular shape. It can be seen that though the ROI mask (it 2, step 4) is edgy and unrefined it can be applied “as is” onto the original object to produce a segmentation without humps (it 2, step 5, above) but if an ellipse is fitted in the ROI mask it produces an Elliptical mask that can segment exactly the main body which is also of a well defined elliptical shape (it 2, step 5, below) inside the original ROI (it 1, step 1). Because the object was ideal the main elliptical body could be segmented precisely without problems producing staggering result for a simple method as it is.

Figure 21. Practical threshold with defect lines



But in actual practice not all the objects are ideal, nonetheless this algorithm works for some of those cases too. Figure 21 Shows an example applying the convex hull with line cuts algorithm in a real case problem. Here the retinal image is over-thresholded to give a full segmentation of the retinal area (It 1, step 1), but as it is seen this not only segmented the retinal area but a part affected by flares and noise produced by the bad focus of the camera plus the effects of the flash light. Because the retina is known to have a round shape it can be assumed that anything attached to it as that big protuberance is not part of it but something else. Applying the convex hull (It 1, step 2) and separating the objects by its defects (It 1, step 3) as in the ideal case previously explained gives as expected the main body of the object as shown in (It 1, step 3 and step 4), which a the end can be used to apply the original produced mask onto the over-threshed object to produce a under-threshed object with a better

segmentation of the retinal area. But as it is known the retina is not formed by straight lines so a smooth version is produced by fitting an ellipse onto the ROI mask to produce the Elliptical mask which gives a more appealing result (it 1, step 5).

Binary masks. There are common methods to threshold images which produce binary mask to use when processing restricted areas in an image. Unfortunately for retinal applications common methods do not work and even the *Otsu* threshold does not work well for the used cases in this project. To overcome this several developed methods were tested but proved inadequate or failed to the task. Below an experimental methods is presented which currently fails in some cases but an additional developed method has been successful so far and performs well in the application with retinal images with severe noise cases. More successful methods are found in `RRtoolbox.lib.arrayops.mask`.

Experimental watershed method to find optic disk area. There is a test that uses watershed to segment bright areas with the objective of obtaining the optic disk and the body of the retina but this proved to be unstable for many cases. The seed points were provided using a histogram analysis to obtain stable points taking into account the background, retinal area and bright areas like flares and the optic disk.

Figure 22. Histogram analysis to find threshold values to use in watershed method

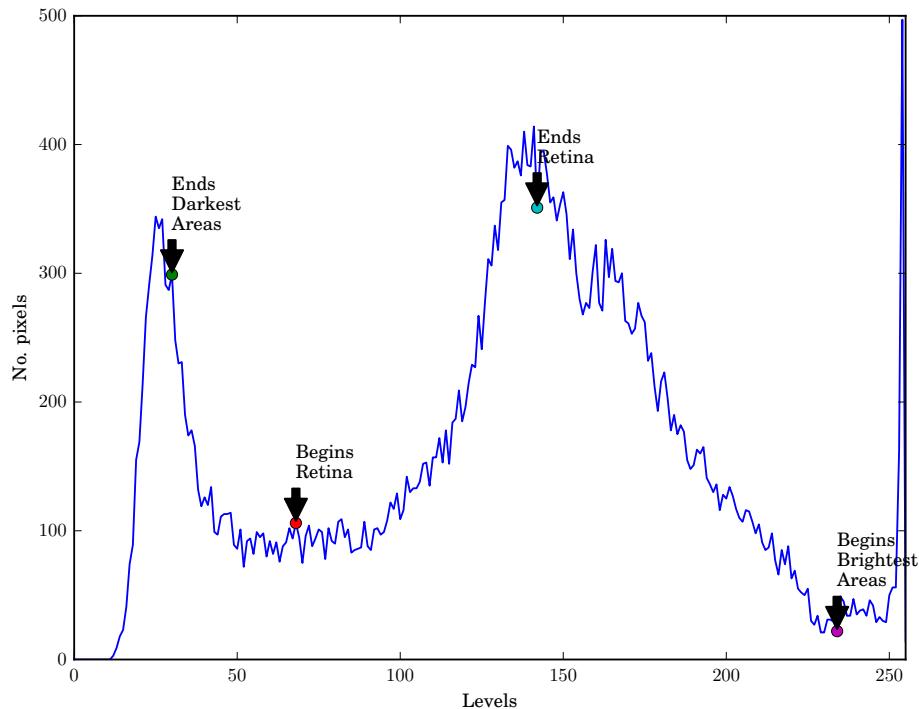
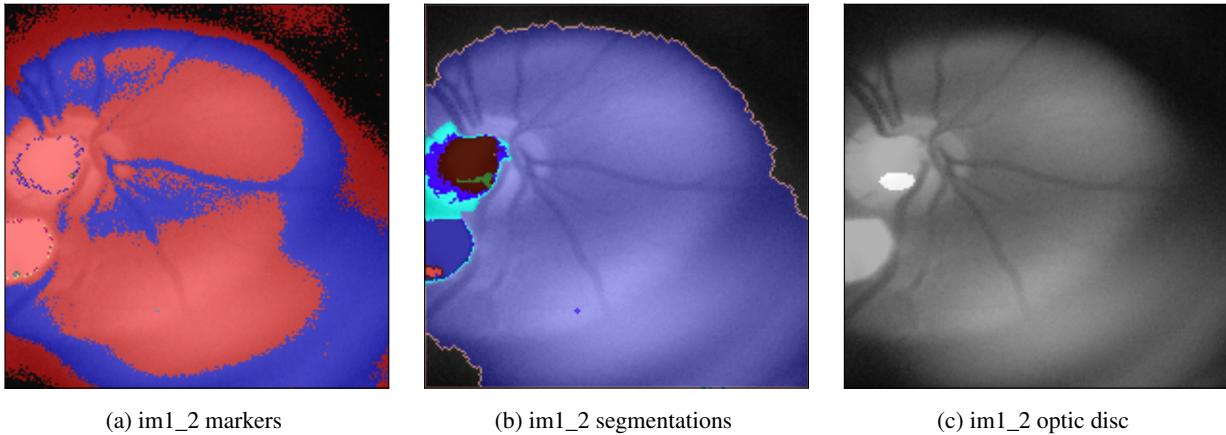


Figure 22 shows a case with the result of the automatic histogram analysis applied to provide the necessary information in the watershed method. The result of feeding these parameters in the watershed method are shown in Figure 23. In Figure 23a the seeds are placed in the sample image which produce the watershed in Figure 23b. It is observed that the watershed not only segmented the retinal area but the blurry area around it. Lastly the brightest areas are analysed to guess what is the optic disk in of these segmentations but as there is a flare present near it the algorithm is confused presenting a foul result trying to not select the flare (Figure 23c).

This algorithm is still in the experimental stage and presents many issues to solve. Both algorithms used to analyse histograms (`retina_markers_thresh`) and apply the watershed method (`find_optic_disc_watershed`) can be imported using `from RRtoolbox.tools.segmentation import retina_markers_thresh, find_optic_disc_watershed` in Python.

Figure 23. Brightest areas in image using watershed method



Proposed threshold method. A method using the *Otsu* threshold was developed to proximate the segmentation to an object at each iteration. The algorithm was implemented in a function called `multiple_otsu` and it was adapted for concrete applications in the background and foreground functions and as their names imply these are used to threshold the background and foreground in an image. These functions can be imported in Python with the `RRtoolbox` package using `from RRtoolbox.lib.arrayops.mask import multiple_otsu, background, foreground`.

Figure 24 shows the comparison between threshold methods. In its column 3 shows the overlaid mask obtained using the background mask over the input image and in column 4 it can be seen the result for the foreground threshold. Wait! but the foreground threshold performs worse than the Otsu method in column 2. Well it is not totally true, the foreground method is just the inverse of the background method so that these two can be compatible. Notice that the background method yields what it promises and produces a better background mask than the Otsu method presented in column 2 and it happens that the foreground method yields the inverse of it.

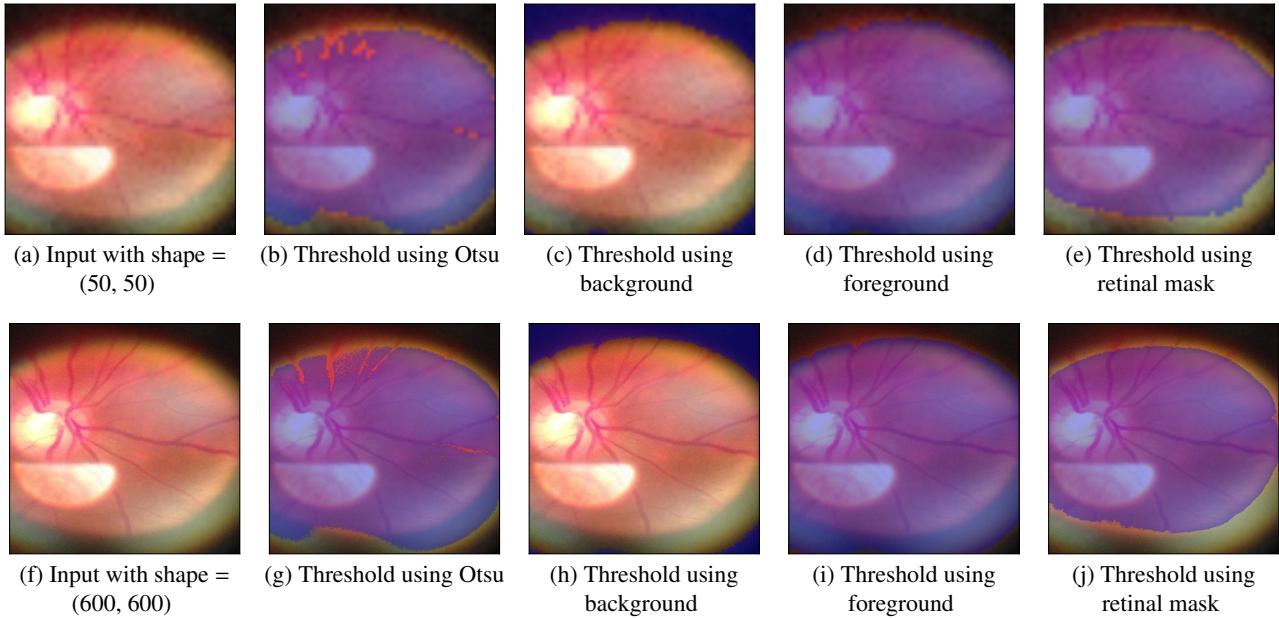
Developed method to find retinal area. There is another developed algorithm that can find ‘solid’ objects in an image without getting confused by blurry areas. This algorithm is called `layeredfloods` because it uses a weighted sum with the flooding algorithm but it has the condition of using color images for better results. If a gray image is fed to the algorithm it tends to yield the same results as the normal Otsu method. This algorithm does not yield exactly a binary mask but an alpha mask which is discussed in more detail in the Alpha masks section . The `layeredfloods` algorithm then was adapted to a new function called `retinal_mask` effectively used to threshold retinal images. Both `layeredfloods` and `retinal_mask` are found in the `RRtoolbox` package and can be imported using `from RRtoolbox.tools.segmentation import layeredfloods, retinal_mask`. The source code of these algorithms are also found there.

Column 5 in Figure 24 shows the overlaid mask over an input image obtained using the `retinal_mask` function. Notice that the algorithm is robust for different conditions like image shape, illumination, color changes (notice how the Otsu method is opened in some areas because the color there was not covered by the threshold value) and blurry areas.

Alpha masks. Though binary masks can accomplish some objectives one begins to realize that they present some limitations as they are exactly designed to segment and nothing more. A better way can be offered using Alpha masks which represent levels of transparency in images. This of course at the expense working with more information, using complex methods to process them and being more resource intensive. Nonetheless, today computers allow to quickly and easily use these masks opening new ways of computing images.

Alpha masks add a layer of information to images by appending an additional dimension to its array. Then BGR images must be transformed to BGRA or RGB to RGBA corresponding to the Red, Green, Blue and Alpha channels used in the image. This is usually used in the PNG image format allowing its images to have transparency when overlaying with others to create composites. A similar example was shown in Figure 6 using the alpha parameter in the overlay algorithm. Next paragraphs explains how to create two alpha masks for the segmentation of retinal images and how these same masks can be used for overlay operations, adding more information to segmentations, adjusting image colors and filtering them.

Figure 24. Spacial filters comparison



Create an alpha mask using filters. The Custom filters using *normSigmoid* section explained the use of the *normSigmoid* in Equation 8 to create custom filters. These filters create a normalized output when fed with an image which can be treated as an *alpha* array and do some useful operations with it. For instance *alpha* can be multiplied (point to point) to the image attenuating it (if *alpha* is normalized this operation does not amplify it).

There are cases where custom filters produce non-normalized results due to adding and subtracting operations. This is solved using Equation 6 from the Normalization section ensuring data integrity (*uint8* data types in *Numpy* are truncated when trying to surpass the 255 value).

Without more further adieu I will explain a simple code that can be imported using `from RRtoolbox.tools.segmentation import get_bright_alpha` which uses filters to create an alpha mask that can be used to merge a foreground and background image. The working example is presented in Code 3 and Figure 25 shows the histogram analysis done by `get_beta_params_hist` to find the beta values and parametrize the filters (blue for background and green for foreground). Additionally the lateral view of the normalized filter response is presented too (in red) obtained with a simulation of the 256 levels in a gray image.

A clearer representation of what the Code 3 does is shown in Figure 26 with the 3D filter response of the simulated 0-255 grey levels of the background and foreground images respectively instead of the actual real images. This demonstrates that the algorithm in Code 3 acts in the color channels and not in the space domain. Notice that if a lateral slice is made across the graph from origin to origin at each axis, that slice corresponds to the merged red plot in Figure 25.

Once the algorithm in Code 3 is used with the real images it creates the alpha mask `normalizedmask` from their gray color channels. This mask can be used with the same background and foreground images to merge them which shown in Figure 28c.

Create an alpha mask for the space domain. There is an studier method using the `layeredfloods` algorithm and it is implemented in the *RRtoolbox* package. This algorithm can be imported using `from RRtoolbox.tools.segmentation import get_layered_alpha`. Figure 27 shows the 3D representation of the alpha mask with row and column axes obtained using `get_layered_alpha`. Notice that the rows and columns indicate that any array operating the alpha mask must have the same dimensions and this mask acts in the space domain. The overlaid result of using this mask is shown in Figure 28d.

Code 3: MWE to create alpha mask with filters

```
# import necessary functions
import cv2
from RRtoolbox.lib.arrayops import normalize, Bandpass, Bandstop
from RRtoolbox.tools.segmentation import get_beta_params_hist

# load images
back = "im1_1.jpg"
fore = "asift2fore.png"
backgray = cv2.imread(back, 0) # load the background image
foregray = cv2.imread(fore, 0) # load the foreground image

# get background beta parameters
beta1B, beta2B = get_beta_params_hist(backgray)

# process backgray in Bandstop filter with automatic parameters
# for example beta1 = 50, beta2 = 190
backmask = Bandstop(alpha = 3, beta1 = beta1B, beta2 = beta2B)(backgray)

# get foreground beta parameters
beta1F, beta2F = get_beta_params_hist(foregray)

# process foregray in Bandpass filter with automatic parameters
# for example beta1 = 50, beta2 = 220
foremask = Bandpass(alpha = 3, beta1 = beta1F, beta2 = beta2F)(foregray)

# merge masks
mergedmask = foremask * backmask

# normalize to [0,1]
normalizedmask = normalize(mergedmask)
```

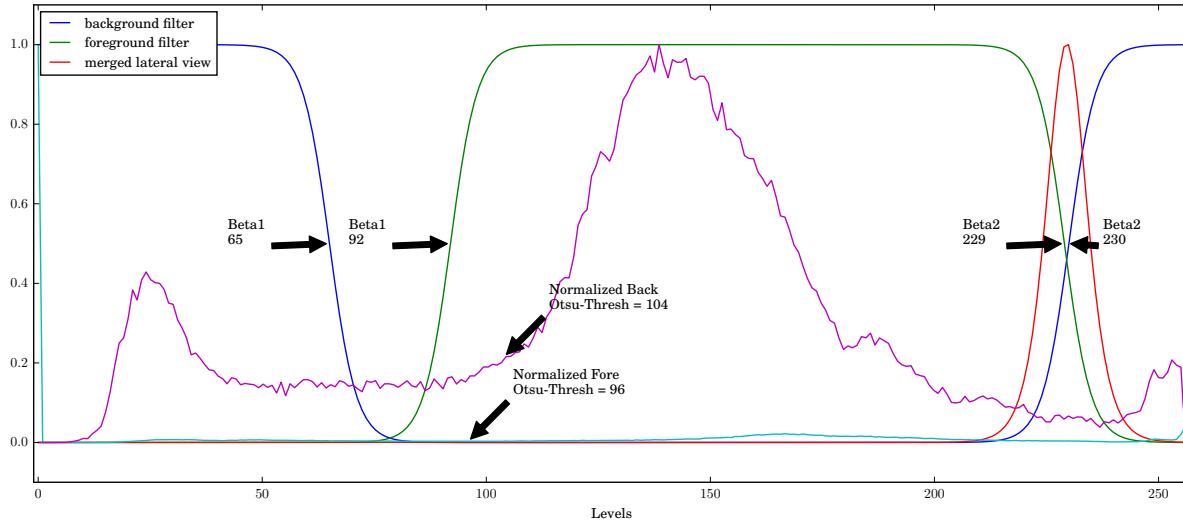
Object Recognition and Matching algorithms

Entropy. It is a classification algorithm from images with multi-focus differences based from (Liu & Yu, 2015). The images must be of the same scene differing only the focus of the camera's lens when taking the picture. They used this algorithm to create high detailed images by combining with the wavelet transform multiple images with diverse focuses into a single focused one. The entropy algorithm based from it can be imported from the *RRtoolbox* package using `from RRtoolbox.tools.selectors import entropy` and Figure 29 shows an example sorting images using this function to classify a set of unfocused images of the same object.

Histogram comparison. The algorithm was implemented using OpenCV function `cv2.compareHist` and based from (Rosebrock, 2014). See the source code and documentation in the repository (David, 2016b, 2016a). Import it for use in Python with `from RRtoolbox.tools.selectors import hist_comp`. Figure 30 shows an application sorting images using the histogram comparison algorithm.

Histogram matching. The purpose of this algorithm is as its name indicates match the histogram of two image to convert the color of one images to the other. This algorithm was conceived trying to recreate `imhistmatch` function from MATLAB (MathWorks, n.d.-a) and based from an stackoverflow answer in (Ali_m, n.d.). Import the created algorithm with `from RRtoolbox.lib.image import hist_match` or see the source code and documentation in the repository (David, 2016b, 2016a). Additional explanations found in (Gonzalez et al., 2004, section 3.3.3). Figure 31 shows an example using the matching algorithm.

Figure 25. Alpha mask histogram analysis and filters response



Detecting and computing features. OpenCV offers a variety of detectors and descriptors (OpenCV, n.d.-a) as listed below:

- *BRISK*, Binary Robust Invariant Scalable Keypoints (Leutenegger, Chli, & Siegwart, 2011).
- *FAST*, Features from Accelerated Segment Test (Rosten & Drummond, 2006; Rosten, Porter, & Drummond, 2010).
- *FREAK*, Fast Retina Keypoint (Alahi, Ortiz, & Vandergheynst, 2012).
- *MSER*, Maximally Stable Extremal Regions (Forssen, 2007; Nistér & Stewénius, 2008).
- *ORB*, Oriented FAST and Rotated BRIEF (Rublee, Rabaud, Konolige, & Bradski, 2011).
- *SIFT*, Scale Invariant Feature Transform (Rey Otero & Delbracio, 2014).
- *SURFT*, Speeded-Up Robust Features (Zhou & Asari, 2011).

SIFT is preferable and selected as the default to apply in the algorithms requiring pattern recognition as it is one of the oldest and successful algorithms for object recognition, image stitching and other applications (Rublee et al., 2011). But because it is licensed and only can be used freely for research purposes (SURFT is licensed too) the other matchers can be used as well.

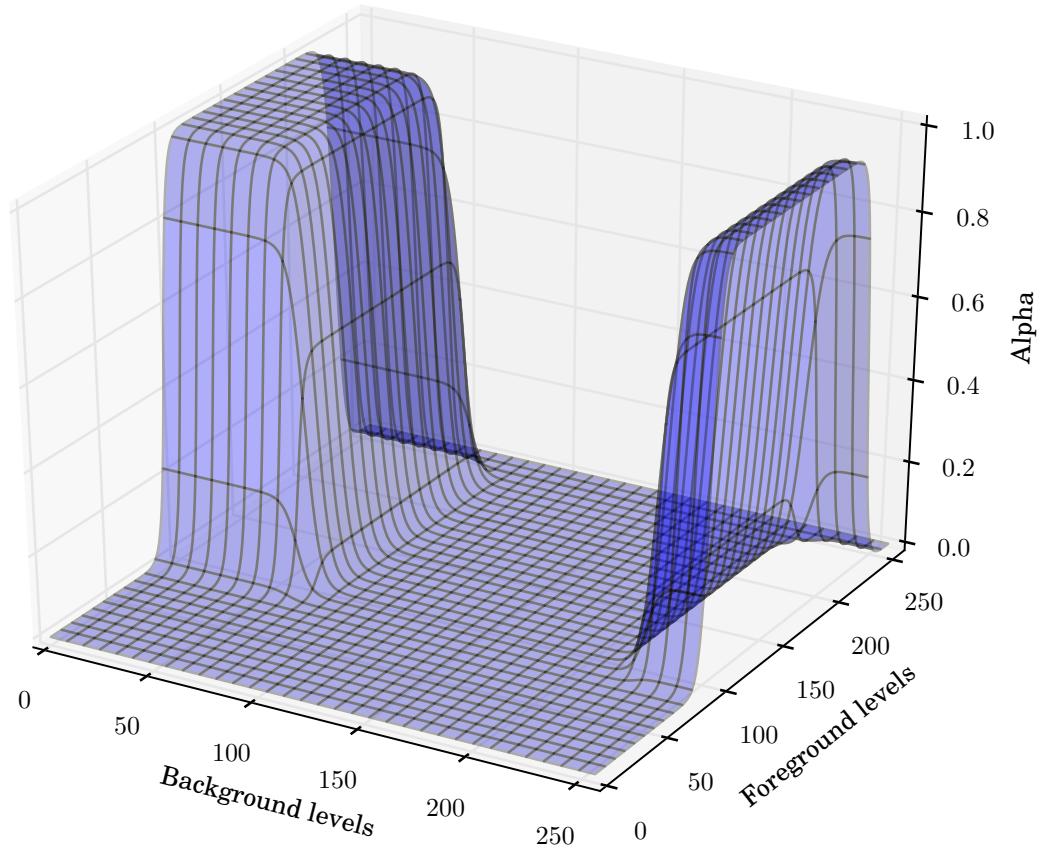
Figure 32 shows a feature detection example using the SIFT algorithm. The colourful circles represent the computed key-points. where the lines from their origin represents the orientation and the radius the size of the key-point. Notice how the rotation and image resolution does not affects much the generation of robust key-points.

ASIFT. ASIFT is a method used to improve on top of a feature-based image matching algorithm by creating various affine transformations of the image that is being computed for its descriptors with the normal matching program (G. Yu & Morel, 2011). The matching algorithm must be scale and rotation invariant to local image features for it to work with ASIFT. The implementation in Python was based from an OpenCV example at (Alexander Alekhin, n.d.) and the code is in the repository with path *RRtoolbox/lib/descriptors.py*. The implementation to use with any matching algorithm provided by OpenCV is the class Feature which can be imported using `from RRtoolbox.lib.descriptors import Feature`.

Using rates and probabilities

There are many ways that results can be rated and that is using probabilities and rates based from robust values. The *RRtoolbox* use them regularly and this section treats some of them. For example the number of descriptors can be used to determine if an image is detailed with many edges, an inlier tests consist of obtaining the rate between inline key-points and number of lines in a homography, rectangularity the measure of how much an object resembles a rectangle, regularity Ratio of

Figure 26. Alpha mask obtained using filters



forms with similar measurements and angles and the convexity ratio the measure of how much an object is convex or resembles its convex hull. The implementation of some of these rates are encapsulated in the `Imcoors` class which can be imported using `from RRtoolbox.lib.image import ImCoors`.

Convexity ratio. One way to determine if an object is concave is to compare its area from the convex hull by finding the ratio between these two (Bozeman & Pilling, 2013, Definition 5). This is what the convexity ratio $r_{convexity}$ does:

$$r_{convexity} = \frac{Area_{object}}{Area_{hull}} \quad (16)$$

where $Area_{object}$ is the area of the object (endogon) and $Area_{hull}$ is the area of its convex hull (exogon) which can be calculated using Equation 1 found in The area under a polygon (`polygonArea`) section . Notice that a convex object will give a convexity ratio of 1 and the more an object is concave the rate will tend to 0. This definition differs a little from other variations (Kindratenko, 1997, Part 2: Shape Analysis, pp. 36–37).

Figure 33 shows an example using the convexity ratio equation. In this example the object points are:

Figure 27. Alpha mask obtained using layered masks

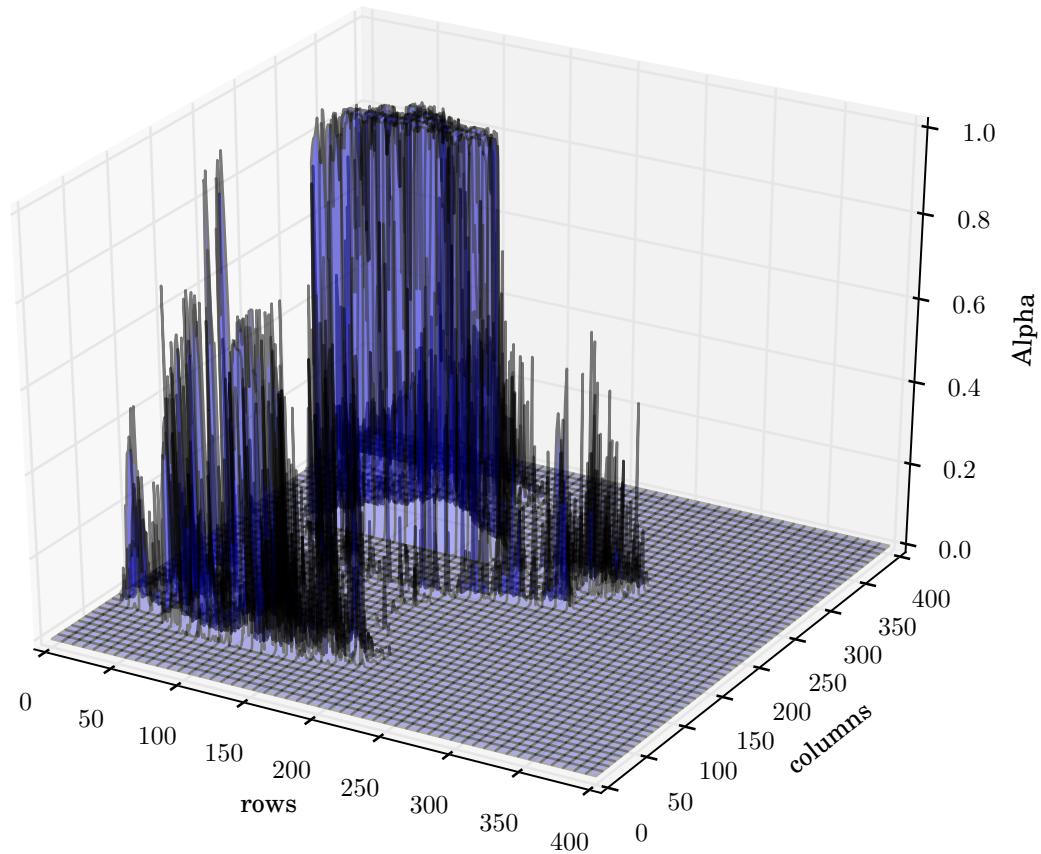
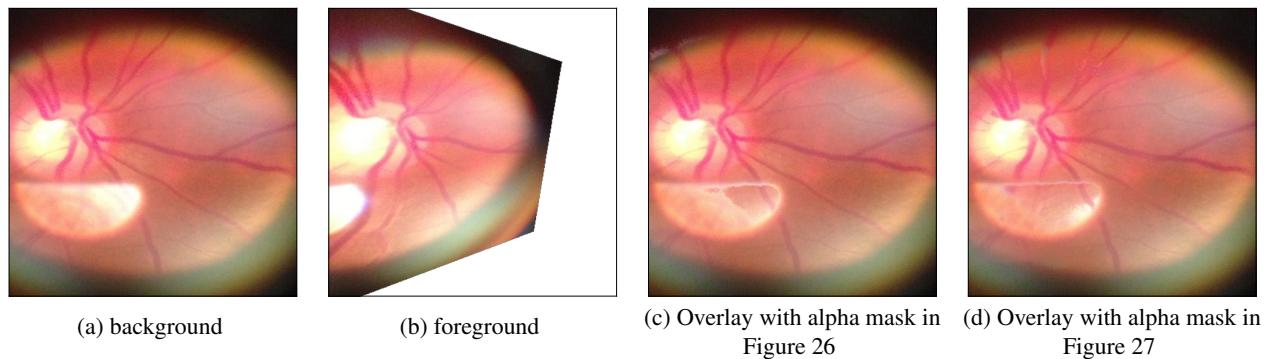
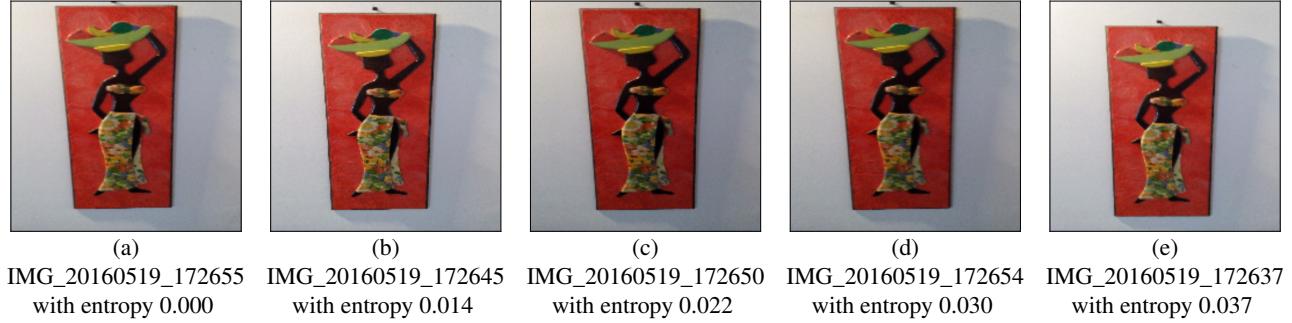


Figure 28. Overlay example using alpha masks



$$object_{points} = \begin{bmatrix} 6.0 & -3.0 \\ 3.0 & 0.0 \\ 5.0 & 4.0 \\ 0.0 & 4.0 \\ -5.0 & 5.0 \\ -2.0 & 0.0 \\ -6.0 & 0.0 \\ -1.0 & -3.0 \\ 0.0 & -6.0 \\ 2.0 & -1.0 \end{bmatrix}$$

Figure 29. Ordered images using entropy



and the points from its Convex hull:

$$hull_{points} = \begin{bmatrix} 6.0 & -3.0 \\ 5.0 & 4.0 \\ -5.0 & 5.0 \\ -6.0 & 0.0 \\ 0.0 & -6.0 \end{bmatrix}$$

then using Equation 16 and Equation 1 yields:

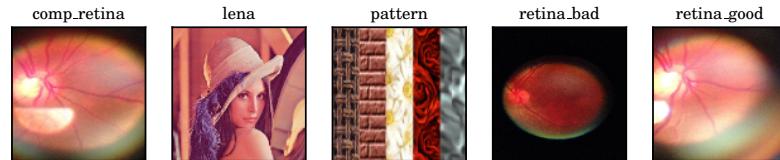
$$\begin{aligned} r_{convexity} &= \frac{Area_{object_{points}}}{Area_{hull_{points}}} \\ &= \frac{53.5}{93.0} \\ &= 0.575268817204 \end{aligned}$$

Rectangularity. This defines the rate to compare how much an object resembles a rectangle. The way to determine if an object resembles a rectangle is to compare its area from the area of a fitted rotated rectangle in the object and finding the ratio between these two.

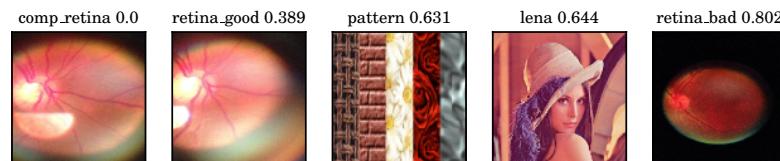
$$rectangularity = \frac{Area_{object}}{Area_{rectangular}} \quad (17)$$

Where $Area_{object}$ is the area of the object and $Area_{rectangular}$ the area of the fitted rotated rectangle that can be provided by OpenCV with the BoxPoints function (see OpenCV documentation for applications). A rectangular object will give a rectangularity of 1 while the rate of an object with a different shape will tend to 0. A simple example is show in Figure 34.

Figure 30. Ordered images using histogram comparison



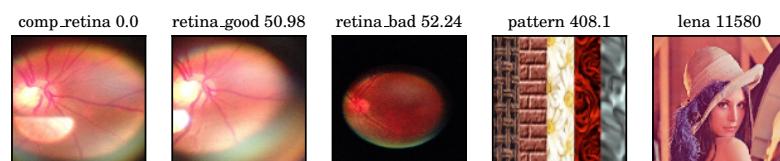
(a) Raw ordered images to sort using histogram comparison



(b) Ordered images using Hellinger



(c) Ordered images using Intersection



(d) Ordered images using Chi-Squared



(e) Ordered images using Correlation

Figure 31. Matching example

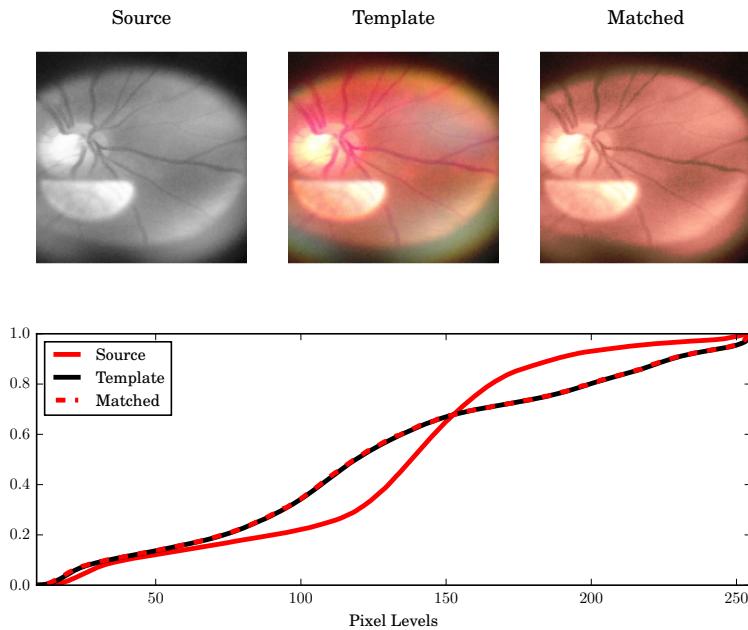


Figure 32. Features using SIFT

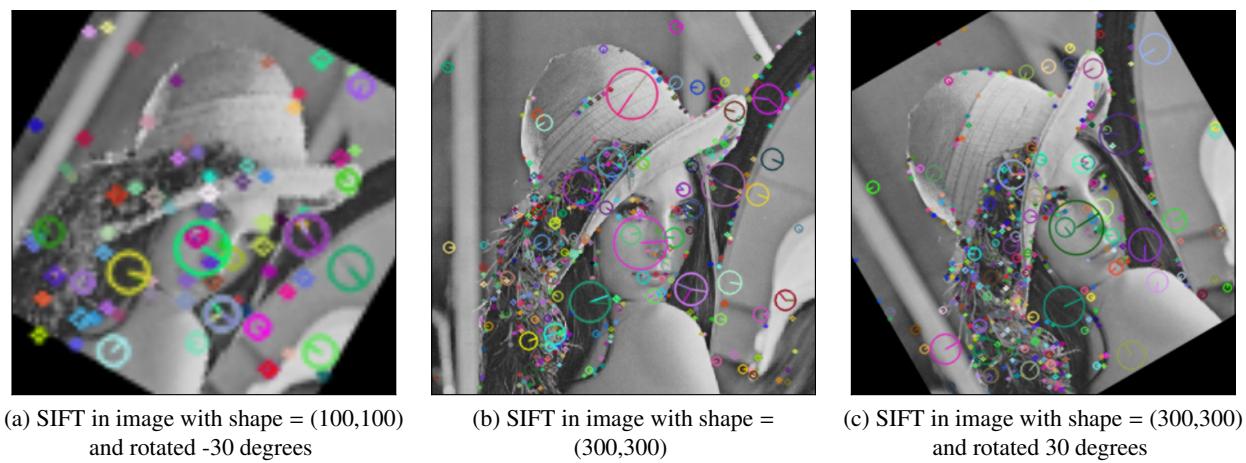
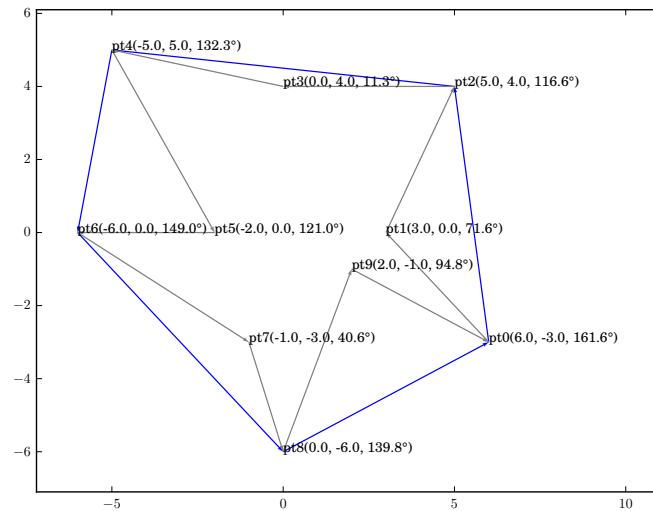
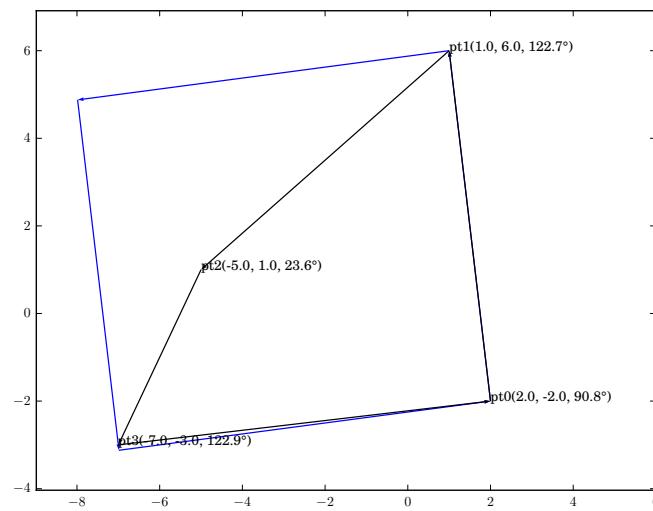


Figure 33. Convexity ratio of an object



(a) ConvexityRatio = 0.575268817204

Figure 34. Rectangularity example



(a) Rectangularity: 0.595890402794

Implementation

The *RRtoolbox* package has many algorithms in development but it already can create some useful applications. This section presents some of them compiled in a complete software called *imrestore* which is an application to restore images in general but in this case is configured to restore retinal images.

The *imrestore* program was developed using a prototyping method which consisted in building step by step the objectives of the restoration tool. Each finished stage of the program was followed by a exhaustive procedure of testing and debugging techniques using profilers, debugging messages, inspection of variables at each line of code, visualization of the processed images, optimizations (both in CPU performance and memory usage) and any needed corrections which leaded to the repetition of the cycle until no further bugs were found. After a stable stage, only then new features were added consisting of additional options to customize the results as well as the coding of the proceeding steps in the program.

Figure 35 shows the class diagram for the main classes of the *imrestore* program. This classes are in the *imrestore.py* script and are called using a *shell* invoked from a terminal or directly from a python environment using the *shell* function to interpret commands as in the terminal, instantiating *ImRestore* for general images or the *RetinalRestore* class for retinal images. The  icon is for classes,  fields,  methods,  deleter properties,  getter properties and  for setter properties.

- Local features: key-points and descriptors:
 - (1.1) SIFT, SURF, ORB, etc.
 - ASIFT*
- Select main or base image from set for merging: raw, sorting, user input
- Matching (spacial):
 - Filter 0.7 below Hamming distance
 - Key points classification
- Selection in matching set: (pre-selection of good matches)
 - Best matches: for general purpose
 - Entropy: used when set is ensured to be of the same object (The program ensures that, if it is not the case).
 - Histogram comparison: use if set contains unwanted perspectives or images that do not correspond to image.
 - Custom function
- Calculate Homography
- Probability tests: (ensures that the matches images correspond to each other)
- Stitching and merging
 - Histogram matching* (color)
 - Segmentation*
 - Alpha mask calculation*
 - Overlay
- Overall filtering*: bilateral filtering
- Lens simulation for retinal photos*

Figure 35. *imrestore.py* classes: *ImRestore* and its inheritor *RetinalRestore*



* optional

Everything related to the code is hosted in (David, 2016b) which contains the main source code package *RRtoolbox*, a development tool using sequential function charts called *RRtoolboxFC* (*FC* stands for Function Chart), documentation, tests, examples and the implemented *imrestore.py* script along with some supporting files. *RRtoolbox* documentation file is in (David, 2016a). The current algorithm needs more refining but has improved over precise segmentation of objects under irregular cases to offer robustness to noise areas like blurred areas,

flares, low brightness, low contrast between actual object and background (normal algorithms need clear distinctions between background and foreground where the histogram is bimodal and not multi-modal and are not prepared for any possible case).

Code 4: Imrestore's script code

```

1 #!/usr/local/bin/python
2 # -*- coding: utf-8 -*-
3 """
4 imrestore (oriented to retinal images):
5
6 Restore images by merging and stitching techniques.
7
8 Optimization techniques:
9     resize to smaller versions*
10
11     memoization*:
12         -persistence
13         -serialization and de-serialization
14         -caching
15
16     multitasking*:
17         -multiprocessing
18         -multithreading
19
20     lazy evaluations:
21         -load on demand
22         -use of weak references
23
24 Memory mapped files*
25
26 STEPS:
27
28 (1) Local features: Key-points and descriptors:
29     -(1.1) SIFT, SURF, ORB, etc
30     -ASIFT*
31
32 (2) Select main or base image from set for merging:
33     -Raw, Sorting, User input
34
35 (3) Matching (spacial):
36     -filter 0.7 below Hamming distance
37     -key points classification
38
39 (4) selection in matching set: (pre selection of good matches)
40     (4.1) Best matches: for general purpose
41     (4.2) Entropy: used when set is ensured to be of the same object
42         (The program ensures that, if it is not the case).
43     (4.3) Histogram comparison: use if set contains unwanted
44         perspectives or images that do not correspond to image.
45     (4.4) Custom function
46
47 (5) Calculate Homography
48
49 (6) Probability tests: (ensures that the matches images
50     correspond to each other)
51
52 (7) Stitching and Merging
53     (7.1) Histogram matching* (color)
```

```

54     (7.2) Segmentation*
55     (7.3) Alpha mask calculation*
56     (7.4) Overlay
57
58 (8) Overall filtering*:
59     Bilateral filtering
60
61 (9) Lens simulation for retinal photos*
62
63 * optional
64
65 Notes:
66
67 Optimization techniques:
68
69     Resize to smaller versions: process smaller versions of the
70     inputs and convert the result back to the original versions.
71     This reduces processing times, standardize the way data is
72     processed (with fixed sizes), lets limited memory to be used,
73     allows to apply in big images without breaking down algorithms
74     that cannot do that.
75
76 Memoization:
77     Persistence: save data to disk for later use.
78
79     serialization and de-serialization: (serialization, in
80     python is referred as pickling) convert live objects into
81     a format that can be recorded; (de-serialization, in python
82     referred as unpickling) it is used to restore serialized
83     data into "live" objects again as if the object was created
84     in program conserving its data from previous sessions.
85
86     Caching: saves the result of a function depending or not in
87     the inputs to compute data once and keep retrieving it from
88     the cached values if asked.
89
90 Multitasking:
91     Multiprocessing: pass tasks to several processes using the
92         computer's cores to achieve concurrency.
93     Multithreading: pass tasks to threads to use "clock-slicing"
94         of a processor to achieve "concurrency".
95
96 Lazy evaluations:
97     load on demand: if data is from an external local file, it is
98         loaded only when it is needed to be computed otherwise it is
99         deleted from memory or cached in cases where it is extensively
100        used. For remote images (e.g. a server, URL) or in an inadequate
101        format, it is downloaded and converted to a numpy format in a
102        temporal local place.
103
104     Use of weak references: in cases where the data is cached or
105         has not been garbage collected, data is retrieved through
106         weak references and if it is needed but has been garbage
107         collected it is loaded again and assigned to the weak reference.
108
109 Memory mapped files:
110     Instantiate an object and keep it not in memory but in a file and
111         access it directly there. Used when memory is limited or data is

```

```

112      too big to fit in memory. Slow downs are negligible for read only
113      mmaped files (i.e. "r") considering the gain in free memory, but
114      it is a real drawback for write operations (i.e. "w", "r+", "w+").
115
116      Selection algorithms:
117          Histogram comparison - used to quickly identify the images that
118              most resemble a target
119          Entropy - used to select the best enfoqued images of the same
120              perspective of an object
121
122      Local features: Key-points and descriptors:
123          ASIFT: used to add a layer of robustness onto other local
124          feature methods to cover all affine transformations. ASIFT
125          was conceived to complete the invariance to transformations
126          offered by SIFT which simulates zoom invariance using gaussian
127          blurring and normalizes rotation and translation. This by
128          simulating a set of views from the initial image, varying the
129          two camera axis orientations: latitude and longitude angles,
130          hence its acronym Affine-SIFT. Whereas SIFT stands for Scale
131          Invariant Feature Transform.
132
133      Matching (spacial):
134          Calculate Homography: Used to find the transformation matrix
135              to overlay a foreground onto a background image.
136
137      Filtering:
138          Bilateral filtering: used to filter noise and make the image
139              colors more uniform (in some cases more cartoonist-like)
140
141      Histogram matching (color): used to approximate the colors from the
142          foreground to the background image.
143
144      Segmentation: detect and individualize the target objects (e.g. optic
145          disk, flares) to further process them or prevent them to be altered.
146
147      Alfa mask calculation: It uses Alfa transparency obtained with sigmoid
148          filters and binary masks from the segmentation to specify where an
149          algorithm should have more effect or no effect at all
150          (i.e. intensity driven).
151
152      Stitching and Merging:
153          This is an application point, where all previous algorithms are
154          combined to stitch images so as to construct an scenery from the
155          parts and merge them if overlapped or even take advantage of these
156          to restore images by completing lacking information or enhancing
157          poorly illuminated parts in the image. A drawback of this is that
158          if not well processed and precise information is given or calculated
159          the result could be if not equal worse than the initial images.
160
161      Lens simulation for retinal photos: As its name implies, it is a
162          post-processing method applied for better appeal of the image
163          depending on the tastes of the user.
164      ...
165      from __future__ import division
166      # TODO install openCV 2.4.12 as described in http://stackoverflow.com/a/37283690/5288758
167      # to solve the error Process finished with exit code 139
168      # UPDATE: openCV 2.4.12 does not solve the error Process finished with exit code 139
169

```

```

170     __author__ = 'Davtoh'
171     # needed for installing executable
172     # program imports
173     import os
174     import cv2
175     import warnings
176     import numpy as np
177     from RRtoolbox.tools.lens import simulateLens
178     from RRtoolbox.lib.config import MANAGER, FLOAT
179     from RRtoolbox.lib.image import hist_match
180     from RRtoolbox.lib.directory import getData, getPath, mkPath, increment_if_exits
181     from RRtoolbox.lib.cache import MemoizedDict, LazyDict
182     from RRtoolbox.lib.image import loadFunc, ImCoors
183     from RRtoolbox.lib.arrayops.mask import brightness, foreground, thresh_biggestCnt
184     from multiprocessing.pool import ThreadPool as Pool
185     from RRtoolbox.tools.selectors import hist_map, hist_comp, entropy
186     from RRtoolbox.tools.segmentation import retinal_mask, get_layered_alpha
187     from RRtoolbox.lib.root import TimeCode, glob, lookinglob, Profiler, VariableNotSettable, globFilter
188     from RRtoolbox.lib.descriptors import Feature, inlineRatio
189     from RRtoolbox.tools.segmentation import get_bright_alpha, Bandpass, Bandstop
190     from RRtoolbox.lib.plotter import MatchExplorer, Plotim, fastplt
191     from RRtoolbox.lib.arrayops.filters import getBilateralParameters
192     from RRtoolbox.lib.arrayops.convert import getS0pointRelation, dict2keyPoint
193     from RRtoolbox.lib.arrayops.basic import getTransformedCorners, transformPoint, \
194         im2shapeFormat, normalize, getOtsuThresh, contours2mask, pad_to_fit_H, overlay
195
196
197     def check_valid(fn):
198         """
199             checks that a file is valid for loading.
200             :param fn: filename
201             :return: True for valid, False for invalid.
202         """
203         test = os.path.isfile(fn)
204         if test and getData(fn)[-2].startswith("_"):
205             return False
206         return test
207
208     class ImRestore(object):
209         """
210             Restore images by merging and stitching techniques.
211
212             :param filenames: list of images or string to path which uses glob filter in path.
213                 Loads image array from path, url, server, string
214                 or directly from numpy array (supports databases)
215             :param debug: (0) flag to print messages and debug data.
216                 0 -> do not print messages.
217                 1 -> print normal messages.
218                 2 -> print normal and debug messages.
219                 3 -> print all messages and show main results.
220                     (consumes significantly more memory).
221                 4 -> print all messages and show all stage results.
222                     (consumes significantly more memory).
223                 5 -> print all messages, show all results and additional data.
224                     (consumes significantly more memory).
225             :param feature: (None) feature instance. It contains the configured
226                 detector and matcher.
227             :param pool: (None) use pool Ex: 4 to use 4 CPUs.

```

```

228 :param cachePath: (None) saves memoization to specified path. This is
229     useful to save some computations and use them in next executions.
230     If True it creates the cache in current path.
231
232     .. warning:: Cached data is not guaranteed to work between different
233         configurations and this can lead to unexpected program
234         behaviour. If a different configuration will be used it
235         is recommended to clear the cache to recompute values.
236 :param clearCache: (0) clear cache flag.
237     * 0 do not clear.
238     * 1 check data integrity of previous session before use
239     * 2 re-compute data but other cache data is left intact.
240     * 3 All CachePath is cleared before use.
241     Notes: using cache can result in unexpected behaviour
242         if some configurations does not match to the cached data.
243 :param loader: (None) custom loader function used to load images.
244     If None it loads the original images in color.
245 :param process_shape: (400,400) process shape, used to load pseudo images
246     to process features and then results are converted to the
247     original images. The smaller the image more memory and speed gain
248     If None it loads the original images to process the features but it
249     can incur to performance penalties if images are too big and RAM
250     memory is scarce.
251 :param load_shape: (None) custom shape used load images which are being merged.
252 :param baseImage: (None) First image to merge to.
253     * None -> takes first image from raw list.
254     * True -> selects image with most features.
255     * Image Name.
256 :param selectMethod: (None) Method to sort images when matching. This
257     way the merging order can be controlled.
258     * (None) Best matches.
259     * Histogram Comparison: Correlation, Chi-squared,
260         Intersection, Hellinger or any method found in hist_map
261     * Entropy.
262     * custom function of the form: rating,fn <- selectMethod(fns)
263 :param distanceThresh: (0.75) filter matches by distance ratio.
264 :param inlineThresh: (0.2) filter homography by inlineratio.
265 :param rectangularityThresh: (0.5) filter homography by rectangularity.
266 :param ransacReprojThreshold: (5.0) maximum allowed reprojection error
267     to treat a point pair as an inlier.
268 :param centric: (False) tries to attach as many images as possible to
269     each matching. It is quicker since it does not have to process
270     too many match computations.
271 :param hist_match: (False) apply histogram matching to foreground
272     image with merge image as template
273 :param grow_scene: If True, allow the restored image to grow in shape if
274     necessary at the merging process.
275 :param expert: Path to an expert database. If provided it will use this data
276     to generate the mask used when merging to the restored image.
277 :param maskforeground:(False)
278     * True, limit features area using foreground mask of input images.
279         This mask is calculated to threshold a well defined object.
280     * Callable, Custom function to produce the foreground image which
281         receives the input gray image and must return the mask image
282         where the keypoints will be processed.
283 :param noisefunc: True to process noisy images or provide function.
284 :param save: (False)
285     * True, saves in path with name _restored_{base_image}

```

```

286         * False, does not save
287         * Image name used to save the restored image.
288 :param overwrite: If True and the destine filename for saving already
289     exists then it is replaced, else a new filename is generated
290     with an index "{filename}_{index}.{extension}"
291 """
292
293 def __init__(self, filenames, **opts):
294     self.profiler = opts.pop("profiler", None)
295     if self.profiler is None:
296         self.profiler = Profiler("ImRestore init")
297
298     self.log_saved = [] # keeps track of last saved file.
299
300     # for debug
301     self.verbosity = opts.pop("verbosity", 1)
302
303 ##### GET IMAGES #####
304 if filenames is None or len(filenames)==0: # if images is empty use demonstration
305     #test = MANAGER["TESTPATH"]
306     #if self.verbosity: print "Looking in DEMO path {}".format(test)
307     #fns = glob(test + "*",check=check_valid)
308     #raise Exception("List of filenames is Empty")
309 elif isinstance(filenames, basestring):
310     # if string assume it is a path
311     if self.verbosity: print "Looking as {}".format(filenames)
312     fns = glob(filenames,check=check_valid)
313 elif not isinstance(filenames, basestring) and \
314     len(filenames) == 1 and "*" in filenames[0]:
315     filenames = filenames[0] # get string
316     if self.verbosity: print "Looking as {}".format(filenames)
317     fns = glob(filenames,check=check_valid)
318 else: # iterator containing data
319     fns = filenames # list file names
320
321     # check images
322 if not len(fns)>1:
323     raise Exception("list of images must be "
324                     "greater than 1, got {}".format(len(fns)))
325
326     # for multiprocessing
327     self.pool = opts.pop("pool", None)
328     if self.pool is not None: # convert pool count to pool class
329         NO_CPU = cv2.popNumberofCPUs()
330         if self.pool <= NO_CPU:
331             self.pool = Pool(processes = self.pool)
332         else:
333             raise Exception("pool of {} exceeds the "
334                             "number of processors {}".format(self.pool,NO_CPU))
335
336     # for features
337     self.feature = opts.pop("feature", None)
338     # init detector and matcher to compute descriptors
339     if self.feature is None:
340         self.feature = Feature(pool=self.pool, debug=self.verbosity)
341         self.feature.config(name='a-sift-flann')
342     else:
343         self.feature.pool = self.pool
344         self.feature.debug = self.verbosity

```

```

344
345     # select method to order images to feed in superposition
346     self.selectMethod = opts.pop("selectMethod",None)
347     best_match_list = ("bestmatches", "best matches")
348     entropy_list = ("entropy",)
349     if callable(self.selectMethod):
350         self._orderValue = 3
351     elif self.selectMethod in hist_map:
352         self._orderValue = 2
353     elif self.selectMethod in entropy_list:
354         self._orderValue = 1
355     elif self.selectMethod in best_match_list or self.selectMethod is None:
356         self._orderValue = 0
357     else:
358         raise Exception("selectMethod {} not recognized".format(self.selectMethod))
359
360     # distance threshold to filter best matches
361     self.distanceThresh = opts.pop("distanceThresh",0.75) # filter ratio
362
363     # threshold for inlineRatio
364     self.inlineThresh = opts.pop("inlineThresh",0.2) # filter ratio
365     # ensures adequate value [0,1]
366     assert self.inlineThresh<=1 and self.inlineThresh>=0
367
368     # threshold for rectangularity
369     self.rectangularityThresh = opts.pop("rectangularityThresh",0.5) # filter ratio
370     # ensures adequate value [0,1]
371     assert self.rectangularityThresh<=1 and self.rectangularityThresh>=0
372
373     # threshold to for RANSAC reprojection
374     self.ransacReprojThreshold = opts.pop("ransacReprojThreshold",5.0)
375
376     self.centric = opts.pop("centric",False) # tries to attach as many images as possible
377     # it is not memory efficient to compute descriptors from big images
378     self.process_shape = opts.pop("process_shape", (400, 400)) # use processing shape
379     self.load_shape = opts.pop("load_shape", None) # shape to load images for merging
380     self.minKps = 3 # minimum len of key-points to find Homography
381     self.histMatch = opts.pop("hist_match",False)
382     self.denoise=opts.pop("denoise", None)
383
384 ##### OPTIMIZATION MEMOIZEDIC #####
385     self.cachePath = opts.pop("cachePath",None)
386     self.clearCache = opts.pop("clearCache",[])
387
388     self.expert = opts.pop("expert",None)
389     if self.expert is not None:
390         self.expert = MemoizedDict(self.expert) # convert path
391
392     # to select base image ahead of any process
393     baseImage = opts.pop("baseImage",None)
394     if isinstance(baseImage,basestring):
395         if baseImage not in fns:
396             base, path, name, ext = getData(baseImage)
397             if not path: # if name is incomplete look for it
398                 base, path, name, ext = getData(fns[0])
399             try: # tries user input
400                 # selected image must be in fns
401                 baseImage = lookingglob(baseImage,

```

```

402                               path= "" .join((base, path)),
403                               filelist=fns, raiseErr=True) #,ext=".*"
404             except Exception as e: # tries to find image based in user input
405                 # generate informative error for the user
406                 try:
407                     # look in the file pattern path
408                     baseImage = lookinglob(baseImage,raiseErr=True) #,ext=".*"
409                     # append new image
410                     fns.append(baseImage)
411             except Exception as e2:
412                 e.args = e.args + e2.args + \
413                         ("A pattern could be '{}'".format("".join((name,".*"))),)
414             raise e
415
416             self.baseImage = baseImage
417
418             if self.verbosity: print "No. images {}...".format(len(fns))
419
420             # assign filenames
421             self.filenames = fns
422
423             # make loader
424             self.loader = opts.pop("loader",None) # BGR loader
425             if self.loader is None: self.loader = loadFunc(1)
426             self._loader_cache = None # keeps last image reference
427             self._loader_params = None # keeps last track of last loading options to reload
428
429             self.save = opts.pop("save",False)
430             self.grow_scene = opts.pop("grow_scene",True)
431             self.maskforeground = opts.pop("maskforeground",False)
432             self.overwrite = opts.pop("overwrite",False)
433
434             # do a check of the options
435             if opts:
436                 raise Exception("Unknown keyword(s) {}".format(opts.keys()))
437
438             # processing variables
439             self._feature_list = None
440             self._feature_dic = None
441             self.used = None
442             self.failed = None
443             self.restored = None
444             self.kps_base,self.desc_base = None,None
445
446             @property
447             def denoise(self):
448                 return self._noisefunc
449             @denoise.setter
450             def denoise(self, value):
451                 if value is False:
452                     value = None
453                 if value is True:
454                     value = "mild"
455                 if value in ("mild","heavy","normal",None) or callable(value):
456                     self._noisefunc = value
457                 else:
458                     raise Exception("denoise '{}' not recognised".format(value))
459             @denoise.deleter

```

```

460 def denoise(self):
461     del self._noisefunc
462
463 @property
464 def feature_list(self):
465     if self._feature_list is None:
466         return self.compute_keypoints()
467     return self._feature_list
468 @feature_list.setter
469 def feature_list(self,value):
470     raise VariableNotSettable("feature_list is not settable")
471 @feature_list.deleter
472 def feature_list(self):
473     self._feature_list = None
474
475 @property
476 def feature_dic(self):
477     if self._feature_dic is None:
478         if self.cachePath is not None:
479             if self.cachePath is True:
480                 self.cachePath = os.path.abspath(".") # MANAGER["TEMPPATH"]
481             if self.cachePath == "{temp}":
482                 self.cachePath = self.cachePath.format(temp=MANAGER["TEMPPATH"])
483             memoized = MemoizedDict(os.path.join(self.cachePath, "descriptors"))
484             if self.verbosity: print "Cache path is in {}".format(memoized._path)
485             self._feature_dic = LazyDict(getter=self.compute_keypoint,
486                                         dictionary=memoized)
487             if self.clearCache==3:
488                 self._feature_dic.clear()
489                 if self.verbosity: print "Cache path cleared"
490             else:
491                 self._feature_dic = LazyDict(getter=self.compute_keypoint)
492             if self.clearCache==1 or self.clearCache==2:
493                 # tell LazyDict to recompute data if key is requested
494                 self._feature_dic.cached = False
495             return self._feature_dic
496 @feature_dic.setter
497 def feature_dic(self,value):
498     self._feature_dic = value
499 @feature_dic.deleter
500 def feature_dic(self):
501     self._feature_dic = None
502
503 def load_image(self, path=None, shape=None):
504     """
505     load image from source
506
507     :param path: filename, url, .npy, server, image in string
508     :param shape: shape to convert image
509     :return: BGR image
510     """
511     params = (path, shape)
512     if self._loader_cache is None or params != self._loader_params:
513         # load new image and cache it
514         img = self.loader(path) # load image
515         if shape is not None:
516             img = cv2.resize(img,shape)
517         self._loader_cache = img # this keeps a reference

```

```

518     self._loader_params = params
519     return img
520   else: # return cached image
521     return self._loader_cache
522
523 def compute_keypoint(self, path):
524   img = self.load_image(path, self.load_shape)
525   lshape = img.shape[:2]
526   try:
527     if self.cachePath is None:
528       point = Profiler(msg=path, tag="cached")
529     else:
530       point = Profiler(msg=path, tag="memoized")
531
532     # compare safely if path is in dictionary, this works for LazyDic,
533     # MemoizeDic or normal dictionaries
534     if path not in self.feature_dic or self.clearCache==2 and path in self.feature_dic:
535       raise KeyError # clears entry from cache
536     kps, desc, pshape = self.feature_dic[path] # thread safe
537     if pshape is None:
538       raise ValueError
539     if self.verbosity: print "{} is cached...".format(path)
540   except (KeyError, ValueError) as e: # not memorized
541     point = Profiler(msg=path, tag="processed")
542     if self.verbosity: print "Processing features for {}...".format(path)
543     if lshape != self.process_shape:
544       img = cv2.resize(img, self.process_shape)
545       img = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
546     # get features
547     if self.maskforeground is None:
548       kps, desc = self.feature.detectAndCompute(img)
549     else:
550       mask = None
551       if callable(self.maskforeground):
552         mask = self.maskforeground(img)
553       if self.maskforeground is True:
554         mask = foreground(img)
555
556       if mask is not None and self.verbosity > 4:
557         try:
558           fastplt.overlay(img.copy(), mask*255, alpha=mask*0.5), block=True,
559           title="{} mask to detect features".format(getData(path)[-2]))
560         except:
561           pass
562
563       kps, desc = self.feature.detectAndCompute(img, mask)
564       pshape = img.shape[:2] # get process shape
565       # to memoize
566       self.feature_dic[path] = kps, desc, pshape
567     # re-scale keypoints to original image
568     if lshape != pshape:
569       # this necessarily does not produce the same result
570       """
571       # METHOD 1: using Transformation Matrix
572       H = getS0pointRelation(process_shape, lshape, True)
573       for kp in kps:
574         kp["pt"] = tuple(cv2.perspectiveTransform(
575           np.array([[kp["pt"]]]), H).reshape(-1, 2)[0])

```

```

576     """
577     # METHOD 2:
578     rx,ry = getS0pointRelation(pshape, lshape)
579     for kp in kps:
580         x,y = kp["pt"]
581         kp["pt"] = x*rx,y*ry
582         kp["path"] = path
583     else:
584         for kp in kps: # be very careful, this should not appear in self.feature_dict
585             kp["path"] = path # add paths to key-points
586     # for profiling individual processing times
587     if self.profiler is not None: self.profiler._close_point(point)
588     return kps, desc, pshape
589
590 def compute_keypoints(self):
591     """
592     Computes key-points from file names.
593
594     :return: self.feature_list
595     """
596     ##### Local features: Key-points and descriptors #####
597     with TimeCode("Computing features...\n", profiler=self.profiler,
598                   profile_point=("Computing features",),
599                   endmsg="Computed feature time was {time}\n",
600                   enableMsg=self.verbosity) as timerK:
601         fns = self.filenames
602         self._feature_list = [] # list of key points and descriptors
603         for index,path in enumerate(fns):
604             if self.verbosity: print "\rFeatures {}/{}...".format(index + 1, len(fns)),
605             kps, desc, pshape = self.compute_keypoint(path)
606             # number of key-points, index, path, key-points, descriptors
607             self._feature_list.append((len(kps),index,path,kps,desc))
608
609     return self._feature_list
610
611 def pre_selection(self):
612     """
613     This method selects the first restored image so that self.restored is initialized
614     with a numpy array and self.used should specify the used image preferably in
615     self.feature_list.
616
617     :return: None
618     """
619     ##### Pre-selection from a set #####
620     baseImage = self.baseImage # baseImage option should not be update
621     # initialization and base image selection
622     if baseImage is None: # select first image as baseImage
623         _,_,baseImage,self.kps_base,self.desc_base = self.feature_list[0]
624     elif isinstance(baseImage,basestring):
625         self.kps_base,self.desc_base,_ = self.feature_dict[baseImage]
626     elif baseImage is True: # sort images
627         self.feature_list.sort(reverse=True) # descendant: from bigger to least
628         # select first for most probable
629         _,_,baseImage,self.kps_base,self.desc_base = self.feature_list[0]
630     else:
631         raise Exception("baseImage must be None, True or String")
632
633     if self.verbosity: print "baseImage is", baseImage

```

```

634     self.used = [baseImage] # select first image path
635     # load first image for merged image
636     self.restored = self.load_image(baseImage, self.load_shape)
637
638     def restore(self):
639         """
640             Restore using file names (self.file_names) with base image (self.baseImage)
641             calculated from self.pre_selection() and other configurations.
642
643         :return: self.restored
644         """
645
646         self.pre_selection()
647         self.failed = [] # registry for failed images
648         fns = self.filenames # in this process fns should not be changed
649         ##### Order set initialization #####
650         if self._orderValue: # obtain comparison with structure (value, path)
651             if self._orderValue == 1: # entropy
652                 comparison = zip(*entropy(fns, loadfunc=loadFunc(1, self.process_shape),
653                                         invert=False)[:2])
654                 if self.verbosity: print "Configured to sort by entropy..."
655             elif self._orderValue == 2: # histogram comparison
656                 comparison = hist_comp(fns, loadfunc=loadFunc(1, self.process_shape),
657                                         method=self.selectMethod)
658                 if self.verbosity:
659                     print "Configured to sort by {}...".format(self.selectMethod)
660             elif self._orderValue == 3:
661                 comparison = self.selectMethod(fns)
662                 if self.verbosity: print "Configured to sort by Custom Function..."
663             else:
664                 raise Exception("DEBUG: orderValue {} does "
665                               "not correspond to {}".format(self._orderValue, self.selectMethod))
666         elif self.verbosity: print "Configured to sort by best matches"
667
668         with TimeCode("Restoring ...\\n", profiler=self.profiler,
669                       profile_point=("Restoring",),
670                       endmsg= "Restoring overall time was {time}\\n",
671                       enableMsg= self.verbosity) as timerR:
672
673             while True:
674                 with TimeCode("Matching ...\\n", profiler=self.profiler,
675                               profile_point=("Matching",),
676                               endmsg= "Matching overall time was {time}\\n",
677                               enableMsg= self.verbosity) as timerM:
678                     ##### remaining keypoints to match #####
679                     # initialize key-point and descriptor base list
680                     kps_remain,desc_remain = [],[]
681                     for _,_,path,kps,desc in self.feature_list:
682                         # append only those which are not in the base image
683                         if path not in self.used:
684                             kps_remain.extend(kps)
685                             desc_remain.extend(desc)
686
686                         if not kps_remain: # if there is not image remaining to stitch break
687                             if self.verbosity:
688                                 if self.failed:
689                                     print "No image remains to merge..."
690                             else:
691                                 print "All images have been merged..."

```

```

692         break
693
694     desc_remain = np.array(desc_remain) # convert descriptors to array
695
696     ##### Matching #####
697     # select only those with good distance (hamming, L1, L2)
698     raw_matches = self.feature.matcher.knnMatch(desc_remain,
699             trainDescriptors = self.desc_base, k = 2) #2
700     # If path=2, it will draw two match-lines for each key-point.
701     classified = {}
702     for m in raw_matches:
703         # filter by Hamming, L1 or L2 distance
704         if m[0].distance < m[1].distance * self.distanceThresh:
705             m = m[0]
706             kp1 = kps_remain[m.queryIdx] # keypoint in query image
707             kp2 = self.kps_base[m.trainIdx] # keypoint in train image
708
709             key = kp1["path"] # ensured that key is not in used
710             if key in classified:
711                 classified[key].append((kp1,kp2))
712             else:
713                 classified[key] = [(kp1,kp2)]
714
715     ##### Order set #####
716     # use only those in classified of histogram or entropy comparison
717     if self._orderValue:
718         ordered = [(val,path) for val, path
719                     in comparison if path in classified]
720     else: # order with best matches
721         ordered = sorted([(len(kps),path)
722                           for path,kps in classified.items()],reverse=True)
723
724
725     with TimeCode("Merging ...\\n",profiler=self.profiler,
726                   profile_point="Merging",),
727                   endmsg= "Merging overall time was {time}\\n",
728                   enableMsg= self.verbosity) as timerH:
729
730         # feed key-points in order according to order set
731         for rank, path in ordered:
732             point = Profiler(msg=path) # profiling point
733             ##### Calculate Homography #####
734             mkp1,mkp2 = zip(*classified[path]) # probably good matches
735             if len(mkp1)>self.minKps and len(mkp2)>self.minKps:
736
737                 # get only key-points
738                 p1 = np.float32([kp["pt"] for kp in mkp1])
739                 p2 = np.float32([kp["pt"] for kp in mkp2])
740                 if self.verbosity > 4:
741                     print 'Calculating Homography for {}...'.format(path)
742
743                 # Calculate homography of fore over back
744                 H, status = cv2.findHomography(p1, p2,
745                                               cv2.RANSAC, self.ransacReprojThreshold)
746             else: # not sufficient key-points
747                 if self.verbosity > 1:
748                     print 'Not enough key-points for {}...'.format(path)
749             H = None

```

```

750
751     # test that there is homography
752     if H is not None: # first test
753         # load fore image
754         fore = self.load_image(path, self.load_shape)
755         h,w = fore.shape[:2] # image shape
756
757         # get corners of fore projection over back
758         projection = getTransformedCorners((h,w),H)
759         c = ImCoors(projection) # class to calculate statistical data
760         lines, inlines = len(status), np.sum(status)
761
762         # ratio to determine how good fore is in back
763         inlineratio = inlineRatio(inlines,lines)
764
765         Test = inlineratio>self.inlineThresh \
766             and c.rotatedRectangularity>self.rectangularityThresh
767
768         text = "inlines/lines: {} / {} = {}, ". \
769             "rectangularity: {}, test: {}".format(
770             inlines, lines, inlineratio, c.rotatedRectangularity,
771             ("failed", "succeeded")[Test])
772
773         if self.verbosity>1: print text
774
775         if self.verbosity > 3: # show matches
776             MatchExplorer("Match " + text, fore,
777                           self.restored, classified[path], status, H)
778
779 ##### probability test #####
780 if Test: # second test
781
782     if self.verbosity>1: print "Test succeeded..."
783     while path in self.failed: # clean path in fail registry
784         try: # race-conditions safe
785             self.failed.remove(path)
786         except ValueError:
787             pass
788
789 ##### merging and stitching #####
790     self.merge(path, H)
791
792     # used for profiling
793     if self.profiler is not None:
794         self.profiler._close_point(point)
795
796     if not self.centric:
797         break
798     else:
799         self.failed.append(path)
800 else:
801     self.failed.append(path)
802
803 # if all classified have failed then end
804 if set(classified.keys()) == set(self.failed):
805     # in the classification there could have been some filtered keys that
806     # did not pass the distance test, these are assigned to the failed list
807     self.failed.extend(list(set(self.used) ^
```

```

808                     set(self.failed) ^
809                     set(self.filenames)))
810
811         if self.verbosity:
812             print "Restoration finished, these images do not fit: "
813             for p in self.failed:
814                 print p
815
816             break
817
818     with TimeCode("Post-processing ...\\n",profiler=self.profiler,
819                   profile_point=("Post-processing",),
820                   endmsg= "Post-processing overall time was {time}\\n",
821                   enableMsg= self.verbosity) as timerP:
822         processed = self.post_process_restoration(self.restored)
823         if processed is not None:
824             self.restored = processed
825
826         # profiling post-processing
827         self.time_postprocessing = timerP.time_end
828
829 ##### Save image #####
830     if self.save:
831         self.save_image()
832
833     return self.restored # return merged image
834
835 def save_image(self, path = None, overwrite = None):
836     """
837     save restored image in path.
838
839     :param path: filename, string to format or path to save image.
840         if path is not a string it would be replaced with the string
841         "{path}restored_{name}{ext}" to format with the formatting
842         "{path}", "{name}" and "{ext}" from the baseImage variable.
843     :param overwrite: If True and the destine filename for saving already
844         exists then it is replaced, else a new filename is generated
845         with an index "{filename}_{index}.{extension}"
846     :return: saved path, status (True for success and False for fail)
847     """
848
849     if path is None:
850         path = self.save
851     if overwrite is None:
852         overwrite = self.overwrite
853
854     bbase, bpath, bname, bext = getData(self.used[0])
855     if isinstance(path,basestring):
856         # format path if user has specified so
857         data = getData(self.save.format(path="".join((bbase, bpath)),
858                         name=bname, ext=bext))
859         # complete any data lacking in path
860         for i,(n,b) in enumerate(zip(data,(bbase, bpath, bname, bext))):
861             if not n: data[i] = b
862
863     else:
864         data = bbase,bpath,"_restored_",bname,bext
865
866     # joint parts to get string
867     fn = "".join(data)
868     mkPath(getPath(fn))

```

```

866     if not overwrite:
867         fn = increment_if_exists(fn)
868
869     if cv2.imwrite(fn, self.restored):
870         if self.verbosity: print "Saved: {}".format(fn)
871         self.log_saved.append(fn)
872         return fn, True
873     else:
874         if self.verbosity: print "{} could not be saved".format(fn)
875         return fn, False
876
877 def merge(self, path, H, shape = None):
878     """
879     Merge image to main restored image.
880
881     :param path: file name to load image
882     :param H: Transformation matrix of image in path over restored image.
883     :param shape: custom shape to load image in path
884     :return: self.restored
885     """
886     alpha = None
887
888     if shape is None:
889         shape = self.load_shape
890
891     fore = self.load_image(path, shape) # load fore image
892
893     if self.histMatch: # apply histogram matching
894         fore = hist_match(fore, self.restored)
895
896     if self.verbosity > 1: print "Merging..."
897
898     # process expert alpha mask if alpha was not provided by the user
899     if self.expert is not None:
900
901         # process _restored_mask if None
902         if not hasattr(self, "_restored_mask"):
903             # from path/name.ext get only name.ext
904             bname = "".join(getData(self.used[-1])[-2:])
905             try:
906                 bdata = self.expert[bname]
907                 bsh = bdata["shape"]
908                 bm_retina = contours2mask(bdata["coors_retina"],bsh)
909                 bm_optic_disc = contours2mask(bdata["coors_optic_disc"],bsh)
910                 bm_defects = contours2mask(bdata["coors_defects"],bsh)
911
912                 self._restored_mask = np.logical_and(np.logical_or(np.logical_not(bm_retina),
913                                                               bm_defects), np.logical_not(bm_optic_disc))
914             except Exception as e:
915                 #exc_type, exc_value, exc_traceback = sys.exc_info()
916                 #lines = traceback.format_exception(exc_type, exc_value, exc_traceback)
917                 warnings.warn("Error using expert {} to create self._restored_mask:\n"
918                               " {}{}".format(bname, type(e), e.args))
919
920         # only if there is a _restored_mask
921         if hasattr(self, "_restored_mask"):
922             fname = "".join(getData(path)[-2:])
923             try:

```

```

924         fdata = self.expert[fname]
925         fsh = fdata["shape"]
926         fm_retina = contours2mask(fdata["coors_retina"],fsh)
927         #fm_otic_disc = contours2mask(fdata["coors_optic_disc"],fsh)
928         fm_defects = contours2mask(fdata["coors_defects"],fsh)
929
930         fmask = np.logical_and(fm_retina,np.logical_not(fm_defects))
931
932         self._restored_mask = maskm = np.logical_and(self._restored_mask,fmask)
933
934         h, w = self.restored.shape[:2]
935         alpha = cv2.warpPerspective(maskm.copy().astype(np.uint8), H, (w, h))
936     except Exception as e:
937         warnings.warn("Error using expert {} to create alpha mask:"
938                         " {}".format(fname,type(e),e.args))
939
940     if alpha is None:
941         # pre process alpha mask
942         alpha = self.pre_process_fore_Mask(self.restored,fore,H)
943
944 ##### SUPERPOSE #####
945
946     # fore on top of back
947     alpha_shape = fore.shape[:2]
948     if self.grow_scene: # this makes the images bigger if possible
949         # fore(x,y)*H = fore(u,v) -> fore(u,v) + back(u,v)
950         ((left,top),(right,bottom)) = pad_to_fit_H(fore.shape, self.restored.shape, H)
951         # moved transformation matrix with pad
952         H_back = FLOAT([[1,0,left],[0,1,top],[0,0,1]]) # in back
953         H_fore = H_back.dot(H) # in fore
954         # need: top_left, bottom_left, top_right, bottom_right
955         h2,w2 = self.restored.shape[:2]
956         w,h = int(left + right + w2),int(top + bottom + h2)
957         # this memory inefficient, image is copied to prevent cross-references
958         self.restored = cv2.warpPerspective(self.restored.copy(), H_back, (w, h))
959         fore = cv2.warpPerspective(fore.copy(), H_fore, (w, h))
960     else: # this keeps back shape
961         H_fore = H
962         H_back = np.eye(3)
963         h, w = self.restored.shape[:2]
964         fore = cv2.warpPerspective(fore.copy(), H_fore, (w, h))
965
966     if alpha is None: # if no pre-processing function for alpha implemented
967         alpha = self.post_process_fore_Mask(self.restored,fore)
968
969     if alpha is None: # create valid mask for stitching
970         alpha = cv2.warpPerspective(np.ones(alpha_shape), H_fore, (w, h))
971
972     if self.verbosity > 3: # show merging result
973         fastplt(alpha, title="alpha mask from {}".format(path),block=True)
974
975     self.restored = overlay(self.restored, fore, alpha) # overlay fore on top of back
976
977     if H_fore is None: # H is not modified use itself
978         H_fore = H
979
980     if self.verbosity > 4: # show merging result
981         Plotim("Last added from {}".format(path), self.restored).show()

```

```

982 ##### update base features #####
983 # make projection to test key-points inside it
984 if self.verbosity > 1: print "Updating key-points..."
985 # fore projection in restored image
986 projection = getTransformedCorners(fore.shape[:2],H_fore)
987 # update key-points
988 newkps, newdesc = [], []
989 for _,_,p,kps,desc in self.feature_list:
990     # append all points in the base image and update their position
991     if p in self.used: # transform points in back
992         for kp,dsc in zip(kps,desc): # kps,desc
993             pt = kp["pt"] # get point
994             if H_back is not None: # update point
995                 pt = tuple(transformPoint(pt,H_back))
996                 kp["pt"] = pt
997             # include only those points outside foreground
998             if cv2.pointPolygonTest(projection, pt, False) == -1:
999                 newkps.append(kp)
1000                 newdesc.append(dsc)
1001
1002 elif p == path: # transform points in fore
1003     # include only those points inside foreground
1004     for kp,dsc in zip(kps,desc): # kps,desc
1005         kp["pt"] = tuple(transformPoint(kp["pt"],H_fore))
1006         newkps.append(kp)
1007         newdesc.append(dsc)
1008
1009 # update self.kps_base and self.desc_base
1010 self.kps_base = newkps
1011 self.desc_base = np.array(newdesc)
1012
1013 if self.verbosity > 4: # show keypoints in merging
1014     Plotim("merged Key-points", # draw key-points in image
1015            cv2.drawKeypoints(
1016                im2shapeFormat(self.restored,self.restored.shape[:2]+(3,)),
1017                [dict2keyPoint(index) for index in self.kps_base],
1018                flags=4, color=(0,0,255))).show()
1019 if self.verbosity: print "This image has been merged: {}".format(path)
1020 self.used.append(path) # update used
1021
1022 return self.restored
1023
1024 def post_process_restoration(self, image):
1025     """
1026     Post-process a merged retinal image.
1027
1028     :param image: retinal image
1029     :return: filtered and with simulated lens
1030     """
1031
1032     if callable(self.denoise):
1033         return self.denoise(image)
1034     # detect how much noise to process and convert it to beta parameters
1035     if self.denoise is not None:
1036         # slower but interactive for heavy noise
1037         # filter using parameters and bilateral filter
1038         params = getBilateralParameters(image.shape, self.denoise)
1039         return cv2.bilateralFilter(image, *params)
1040
1041 def post_process_fore_Mask(self, back, fore):
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1041      """
1042      Method to post-process fore mask used after fore and back are transformed
1043      to a new domain for merging. This method is called by the merge method in the
1044      event that an alpha mask has not been created by self.pre_process_fore_Mask
1045      method or obtained with the self.expert variables.
1046
1047      :param back: background image. This is called by method self.merge
1048          with self.restored
1049      :param fore: fore ground image.
1050      :return: alpha mask with shape (None,None)
1051      """
1052      pass
1053
1054  def pre_process_fore_Mask(self, back, fore, H):
1055      """
1056      Method to pre-process fore mask used before fore and back are transformed
1057      to a new domain for merging.
1058
1059      :param back:
1060      :param fore:
1061      :param H:
1062      :return: alpha mask with shape (None,None)
1063      """
1064      pass
1065
1066  class RetinalRestore(ImRestore):
1067      """
1068      Restore retinal images by merging and stitching techniques. These parameters are
1069      added to :class:`ImRestore`:
1070
1071      :param lens: flag to determine if lens are applied. True
1072          to simulate lens, False to not apply lens.
1073      :param enclose: flag to enclose and return only retinal area.
1074          True to return ROI, false to leave image "as is".
1075      """
1076      def __init__(self, filenames, **opts):
1077          # overwrite variables
1078          opts["denoise"] = opts.pop("denoise", True)
1079          opts["maskforeground"] = opts.pop("maskforeground", lambda img: retinal_mask(img, True))
1080          # create new variables
1081          self.lens = opts.pop("lens", False)
1082          self.enclose = opts.pop("enclose", False)
1083          # call super class
1084          super(RetinalRestore, self).__init__(filenames, **opts)
1085
1086      #__init__.__doc__ = ImRestore.__init__.__doc__+__init__.__doc__
1087
1088  def post_process_fore_Mask(self, back, fore):
1089      """
1090      Function evaluated just after superposition and before overlay.
1091
1092      :param back: background image
1093      :param fore: foreground image
1094      :return: alpha mask
1095      """
1096
1097      def _mask(back, fore):
1098          """

```

```

1098     get bright alpha mask (using histogram method)
1099
1100     :param back: BGR background image
1101     :param fore: BGR foreground image
1102     :return: alpha mask
1103     """
1104
1105     # TODO, not working for every retinal scenario
1106     foregray = brightness(fore)
1107     # get window with Otsu to prevent expansion
1108     thresh,w = cv2.threshold(foregray,0,1,cv2.THRESH_BINARY+cv2.THRESH_OTSU)
1109     return get_bright_alpha(brightness(back).astype(float),
1110                             foregray.astype(float), window=w)
1111
1112
1113     pshape = (400,400) # process shape
1114     # rescaling of the image to process mask
1115     if pshape is not None:
1116         oshape = back.shape[:2]
1117         back = cv2.resize(back,pshape)
1118         fore = cv2.resize(fore,pshape)
1119
1120     # get alpha mask
1121     alphamask = get_layered_alpha(back,fore)
1122
1123     # rescaling mask to original shape
1124     if pshape is not None:
1125         alphamask = cv2.resize(alphamask,oshape[::-1])
1126
1127     return alphamask
1128
1129 def post_process_restoration(self, image):
1130     """
1131     Post-process a merged retinal image.
1132
1133     :param image: retinal image
1134     :return: filtered and with simulated lens
1135     """
1136     image_ = super(RetinalRestore,self).post_process_restoration(image)
1137     if image_ is not None:
1138         image = image_
1139
1140     # simulation of lens
1141     if self.lens:
1142         # call function to overlay lens on image
1143         try:
1144             image = simulateLens(image)
1145         except Exception as e:
1146             warnings.warn("simulate lens failed with error {}".format(e))
1147
1148     # crop retinal area only
1149     if self.enclose:
1150         # convert to gray
1151         gray = brightness(image)
1152         # get object mask
1153         mask = foreground(gray)
1154         # get contour of biggest area
1155         cnt = thresh_biggestCnt(mask)
1156         # get enclosure box

```

```

1156     x,y,w,h = cv2.boundingRect(cnt)
1157     # crop image
1158     if len(image.shape)>2:
1159         image = image[x:x+w,y:y+h,:]
1160     else:
1161         image = image[x:x+w,y:y+h]
1162
1163     return image
1164
1165 def feature_creator(string):
1166     """
1167     Converts a string to a feature object.
1168
1169     :param string: any supported feature detector in openCV. the format is
1170         "[a-]<sift/surf/orb>[-flann]" (str) Ex: "a-sift-flann" where
1171         "a-" or "-flann" are optional.
1172     :return: feature object
1173     """
1174     return Feature().config(string)
1175
1176 def tuple_creator(string):
1177     """
1178     Process string to get tuple.
1179
1180     :param string: string parameters with ":" (colon) as separator
1181         Ex: param1,param2,...,paramN
1182     :return: tuple
1183     """
1184     tp = []
1185     func = string_interpreter()
1186     for i in string.split(","):
1187         try:
1188             tp.append(func(i))
1189         except:
1190             tp.append(i)
1191     return tuple(tp)
1192
1193 def loader_creator(string):
1194     """
1195     creates an image loader.
1196
1197     :param string: flag, x size, y size. Ex 1: "0,100,100" loads gray images of shape
1198         (100,100) in gray scale. Ex 2: "1" loads images in BGR color and with
1199         original shapes. Ex 3: "0,200,None" loads gray images of shape (200,None)
1200         where None is calculated to keep image ratio.
1201     :return: loader
1202     """
1203     params = tuple_creator(string)
1204     try:
1205         flag = params[0]
1206     except:
1207         flag=1
1208     try:
1209         x = params[1]
1210     except:
1211         x=None
1212     try:
1213         y = params[2]

```

```

1214     except:
1215         y=None
1216     return loadFunc(flag,(x,y))
1217
1218 def denoise_creator(string):
1219     """
1220     creates an function to de-noise images using bilateral filter.
1221
1222     :param string: d, sigmaColor, sigmaSpace. Ex: 27,75,75 creates the
1223         filter to de-noise images.
1224     :return: denoiser
1225     """
1226     d, sigmaColor, sigmaSpace = tuple_creator(string)
1227     def denoiser(image):
1228         return cv2.bilateralFilter(image, d, sigmaColor, sigmaSpace)
1229     return denoiser
1230
1231 def string_interpreter(empty=None, commahandler=None, handle=None):
1232     """
1233     create a string interpreter
1234     :param empty: (None) variable to handle empty strings
1235     :param commahandler: (tuple_creator) function to handle comma separated strings
1236     :return: interpreter function
1237     """
1238     def interprete_string(string):
1239         if string == "": # in argparse this does not applies
1240             return empty
1241         if "," in string:
1242             if commahandler is None:
1243                 return tuple_creator(string)
1244             else:
1245                 return commahandler(string)
1246         if string.lower() == "none":
1247             return None
1248         if string.lower() == "true":
1249             return True
1250         if string.lower() == "false":
1251             return False
1252         if handle is None:
1253             try:
1254                 return int(string)
1255             except:
1256                 return string
1257         else:
1258             return handle(string)
1259     interprete_string.__doc__="""
1260     Interpret strings.
1261
1262     :param string: string to interpret.
1263     :return: interpreted string. If empty string (i.e. '') it returns {}.
1264         If 'None' returns None. If 'True' returns True. If 'False' returns False.
1265         If comma separated it applies {} else applies {}.
1266     """.format(empty, commahandler, handle)
1267     return interprete_string
1268
1269 class NameSpace(object):
1270     """
1271     used to store variables

```

```

1272 """
1273
1274 def shell(args=None, namespace=None):
1275     """
1276     Shell to run in terminal the imrestore program
1277
1278     :param args: (None) list of arguments. If None it captures the
1279         arguments in sys.
1280     :param namespace: (None) namespace to place variables. If None
1281         it creates a namespace.
1282     :return: namespace
1283     """
1284
1285     if namespace is None:
1286         namespace = NameSpace()
1287
1288     import argparse
1289
1290     parser = argparse.ArgumentParser(formatter_class=argparse.RawDescriptionHelpFormatter,
1291                                     description="Restore images by merging and stitching "
1292                                     "techniques.",
1293                                     epilog=__doc__ +
1294                                         "\nContributions and bug reports are appreciated."
1295                                         "\nauthor: David Toro"
1296                                         "\ne-mail: davsamirtor@gmail.com"
1297                                         "\nproject: https://github.com/davtoh/RRtools")
1298
1299     parser.add_argument('filenames', nargs='*',
1300                         help='List of images or path to images. Glob operations can be '
1301                         'achieved using the wildcard sign "*". '
1302                         'It can load image from files, urls, servers, strings'
1303                         'or directly from numpy arrays (supports databases)'
1304                         'Because the shell process wildcards before it gets '
1305                         'to the parser it creates a list of filtered files in '
1306                         'the path. Use quotes in shell to prevent this behaviour '
1307                         'an let the restorer do it instead e.g. "/path/to/images/*.jpg". '
1308                         'if "*" is used then folders and filenames that start with an '
1309                         'underscore "_" are ignored by the restorer')
1310
1311     parser.add_argument('-v', '--verbosity', type=int, default=1,
1312                         help="""(0) flag to print messages and debug data.
1313                             0 -> do not print messages.
1314                             1 -> print normal messages.
1315                             2 -> print normal and debug messages.
1316                             3 -> print all messages and show main results.
1317                                 (consumes significantly more memory).
1318                             4 -> print all messages and show all results.
1319                                 (consumes significantly more memory).
1320                             5 -> print all messages, show all results and additional data.
1321                                 (consumes significantly more memory).""")
1322
1323     parser.add_argument('-f', '--feature', type=string_interpreter(commahandler=feature_creator),
1324                         help='Configure detector and matcher')
1325
1326     parser.add_argument('-u', '--pool', action='store', type=int,
1327                         help='Use pool Ex: 4 to use 4 CPUs')
1328
1329     parser.add_argument('-c', '--cachePath', default=None,
1330                         help="""
1331                             saves memoization to specified path. This is useful to save
1332                             some computations and use them in next executions.
1333                             Cached data is not guaranteed to work between different
1334                             configurations and this can lead to unexpected program
1335                             behaviour. If a different configuration will be used it
1336                             is recommended to clear the cache to recompute values.
1337                             If True it creates the cache in current path.
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1330
1331     """)
1332     parser.add_argument('-e', '--clearCache', type=int, default=0,
1333                         help='clear cache flag.'
1334                         '/* 0 do not clear.'
1335                         '/* 1 check data integrity of previous session before use'
1336                         '/* 2 re-compute data but other cache data is left intact.'
1337                         '/* 3 All CachePath is cleared before use.'
1338                         'Notes: using cache can result in unexpected behaviour '
1339                         'if some configurations does not match to the cached data.')
1340     parser.add_argument('--loader', type=string_interpreter(commahandler=loader_creator),
1341                         nargs='?', help='Custom loader function used to load images. '
1342                         'By default or if --loader flag is empty it loads the '
1343                         'original images in color. The format is "--loader colorflag, '
1344                         'x, y" where colorflag is -1,0,1 for BGRA, gray and BGR images '
1345                         'and the load shape are represented by x and y. '
1346                         'Ex 1: "0,100,100" loads gray images of shape (100,100) in '
1347                         'gray scale. Ex 2: "1" loads images in BGR color and with '
1348                         'original shapes. Ex 3: "0,200,None" loads gray images of shape '
1349                         '(200,None) where None is calculated to keep image ratio.')
1350     parser.add_argument('-p', '--process_shape', default=(400,400), type=string_interpreter(),
1351                         nargs='?', help='Process shape used to convert to pseudo images '
1352                         'to process features and then convert to the '
1353                         'original images. The smaller the image more memory and speed '
1354                         'gain. By default process_shape is 400,400'
1355                         'If the -p flag is empty it loads the original '
1356                         'images to process the features but it can incur to performance '
1357                         'penalties if images are too big and RAM memory is scarce')
1358     parser.add_argument('-l', '--load_shape', default=None, type=string_interpreter(),
1359                         nargs='?', help='shape used to load images which are being merged.')
1360     parser.add_argument('-b', '--baseImage', default=True, type=string_interpreter(), nargs='?',
1361                         help='Specify image's name to use from path as first image to merge '
1362                         'in the empty restored image. By default it selects the image '
1363                         'with most features. If the -b flag is empty it selects the '
1364                         'first image in filenames as base image')
1365     parser.add_argument('-m', '--selectMethod',
1366                         help='Method to sort images when matching. This '
1367                         'way the merging order can be controlled.'
1368                         '/* (None) Best matches'
1369                         '/* Histogram Comparison: Correlation, Chi-squared, '
1370                         'Intersection, Hellinger or any method found in hist_map'
1371                         '/* Entropy'
1372                         '/* custom function of the form: rating,fn <- selectMethod(fns)')
1373     parser.add_argument('-d', '--distanceThresh', type = float, default=0.75,
1374                         help='Filter matches by distance ratio')
1375     parser.add_argument('-i', '--inlineThresh', type = float, default=0.2,
1376                         help='Filter homography by inlineratio')
1377     parser.add_argument('-r', '--rectangularityThresh', type = float, default=0.5,
1378                         help='Filter homography by rectangularity')
1379     parser.add_argument('-j', '--ransacReprojThreshold', type = float, default=10.0,
1380                         help='Maximum allowed reprojection error '
1381                         'to treat a point pair as an inlier')
1382     parser.add_argument('-n', '--centric', action='store_true',
1383                         help='Tries to attach as many images as possible to '
1384                         'each matching. It is quicker since it does not have to process '
1385                         'too many match computations')
1386     parser.add_argument('-t', '--hist_match', action='store_true',
1387                         help='Apply histogram matching to foreground '
1388                         'image with merge image as template')

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1388 parser.add_argument('-s', '--save', default=True, nargs='?', action="store",
1389     const=False, type = string_interpreter(False),
1390     help='Customize image name used to save the restored image.'
1391         'By default it saves in path with name "_restored_{base_image}".'
1392         'if the -s flag is specified empty it does not save. Formatting '
1393         'is supported so for example the default name can be achived as '
1394         '"-s {path}_restored_{name}{ext}"')
1395 parser.add_argument('-o', '--overwrite', action='store_true',
1396     help = 'If True and the destine filename for saving already '
1397         'exists then it is replaced, else a new filename is generated '
1398         'with an index "{filename}_{index}.{extension}"')
1399 parser.add_argument('-g', '--grow_scene', action='store_true',
1400     help='Flag to allow image to grow the scene so that that the final '
1401         'image can be larger than the base image')
1402 parser.add_argument('-y', '--denoise', nargs='?', action="store", const=True,
1403     type=string_interpreter(True, commahandler=denoise_creator),
1404     help="Flag to process noisy images. Use mild, normal, heavy or "
1405         "provide parameters for a bilateral filter as "
1406         "'--denoise d,sigmaColor,sigmaSpace' as for example "
1407         "'--denoise 27,75,75'. By default it is None which can be "
1408         "activated according to the restorer, if an empty flag is "
1409         "provided as '--denoise' it deactivates de-noising images.")
1410 parser.add_argument('-a', '--lens', action='store_true',
1411     help='Flag to apply lens to retinal area. Else do not apply lens')
1412 parser.add_argument('-k', '--enclose', action='store_true',
1413     help='Flag to enclose and return only retinal area. '
1414         'Else leaves image "as is"')
1415 parser.add_argument('-z', '--restorer', choices = ['RetinalRestore', 'ImRestore'],
1416     default='RetinalRestore',
1417     help='imrestore is for images in general but it can be parametrized. '
1418         'By default it has the profile "retinalRestore" for retinal '
1419         'images but its general behaviour can be restorerd by '
1420         'changing it to "imrestore")')
1421 parser.add_argument('-x', '--expert', default=None, help='path to the expert variables')
1422 parser.add_argument('-q', '--console', action='store_true',
1423     help='Enter interactive mode to let user execute commands in console')
1424 parser.add_argument('-w', '--debug', action='store_true', # https://pymotw.com/2/pdb/
1425     help='Enter debug mode to let programmers find bugs. In the debugger '
1426         'type "h" for help and know the supported commands.')
1427 parser.add_argument('--onlykeys', action='store_true',
1428     help='Only compute keypoints. This is useful when --cachePath is '
1429         'used and the user wants to have the keypoints cached beforehand')
1430
1431 # parse sys and get argument variables
1432 args = vars(parser.parse_args(args=args, namespace=namespace))
1433
1434 # shell variables
1435 debug = args.pop('debug')
1436 console = args.pop('console')
1437 onlykeys = args.pop('onlykeys')
1438
1439 # debugger
1440 if debug:
1441     print "debug ON."
1442     import pdb; pdb.set_trace()
1443
1444 # this is needed because the shell process wildcards before it gets to argparse
1445 # creating a list in the path thus it must be filtered. Use quotes in shell

```

```

1446 # to prevent this behaviour
1447 if len(args['filenames'])>1:
1448     args['filenames'] = [p for p in args['filenames'] if check_valid(p) or "*" in p]
1449 else:
1450     args['filenames'] = args['filenames'][0]
1451
1452 # print parsed arguments
1453 if args['verbosity']>1:
1454     print "Parsed Arguments\n",args
1455
1456 # use configuration
1457 use_restorer = args.pop("restorer")
1458 if use_restorer == 'RetinalRestore':
1459     if args['verbosity']: print "Configured for retinal restoration..."
1460     self = RetinalRestore(**args)
1461 elif use_restorer == 'ImRestore':
1462     if args['verbosity']: print "Configured for general restoration..."
1463     for key in ['enclose', 'lens']:
1464         args.pop(key) # clean up unused key
1465     self = ImRestore(**args)
1466 else:
1467     raise Exception("no restoration class called {}".format(use_restorer))
1468
1469 if namespace is not None:
1470     # update namespace from early stages so it can have access to the restorer
1471     namespace.restorer = self
1472
1473 if console:
1474     print "interactive ON."
1475     print "restoring instance is 'namespace.restorer' or 'self'"
1476     print "Ex: type 'self.restore()' to proceed with restoration."
1477     import code; code.interact(local=locals())
1478 elif onlykeys:
1479     self.compute_keypoints()
1480 else:
1481     # start restoration
1482     self.restore()
1483
1484 return namespace # return given namespace
1485
1486 if __name__ == "__main__":
1487     shell() # run the shell
1488     # TODO visualizer for alpha mask
1489     # TODO implement standard deviation in bright areas to detect optic disk

```

After the script is completed the creation of the executable can be done using a shell interface or a graphical interface can be added on top of it. This is further explained in User Interface section .

User Interface

This section gives an introduction to the *imrestore* interfaces which can be resumed in its classes, command-line interface using a shell and the developing GUI more detailed discussed in Future Work.

Command-line interface

Usually the following command format is used to run the *imrestore* script:

```
python imrestore.py <path to folder with images>
```

`python imrestore.py` calls Python interpreter to run `imrestore.py` script but if the file is a bash or executable `./<path to executable>` is used. For example, to test `imrestore` once Pyinstaller finishes creating the executable as explained in subsection *Executable* just enter the command:

```
./dist/imrestore tests/im1*
```

The `.` is used to signify that the operation is from respect the current path. So if the terminal's path is in `dist` folder use:

```
./imrestore tests/im1*
```

For further reading the capabilities of `imrestore` and configurations just type in:

```
./imrestore -h
```

or its equivalent

```
./imrestore --help
```

Executable

There are several options to create executables and installers from python sources (“deployment - Python Wiki,” n.d.). From them Pyinstaller suffices for all the needs to create them. It is a python package so it must be installed after python, these procedures are found in subsubsection A.1.3 and further reading in (Cortesi, Bajo, Caban, & McMillan, n.d.). To create one-file executable from a script the following command format is used:

```
pyinstaller -p <path to sources> -n <name of program> -F <path to script> --version-file=<path to version file>
```

The `-p` option is to add a path to search for imports, without this the executable would not have all the features. `-n` Name to assign to the bundled *app* and *spec* file. `-F` Create a one-file bundled executable.

Lets create `imrestore` executable then, so open the Terminal where `RRtoolbox` folder and `imrestore.py` are (i.e. like in the repository) then type the following command to create the executable:

```
pyinstaller -p ./ -n imrestore -F ./imrestore.py --version-file=version
```

This will create the folders *build* where temporal files are compiled to be used in the creation of the executable and *dist* where the executable is placed once pyinstaller finishes building it. Under windows 7 the `msvcr100.dll` is needed so it is advised to install Microsoft Visual C++ 2010 before running pyinstaller to let it find the necessary files. Notice that the executable is only compatible for computers with the same operating system as the one used to create it.

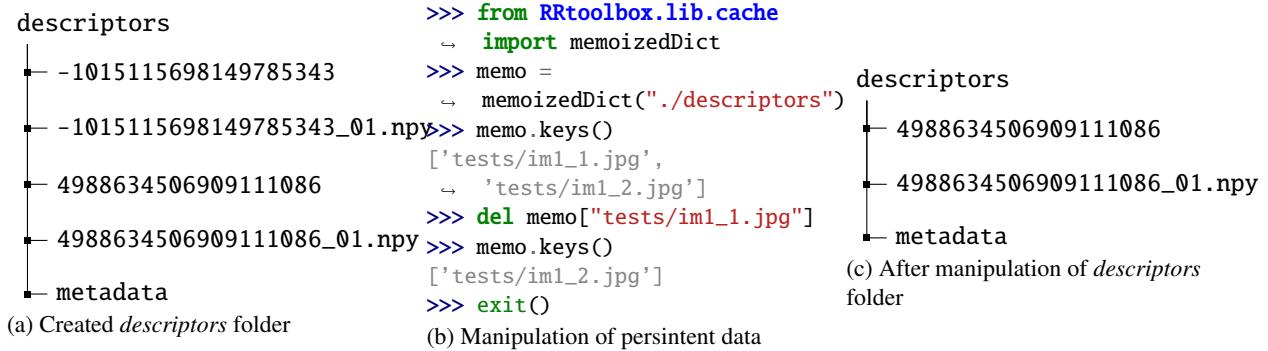
Results

The majority of results were taken using `imrestore.py` script with generic options (see Code 5) where samples represented worse-case scenarios that could be obtained in a non-professional way, ensuring that a professional ophthalmologist can obtain better results than the ones here presented.

To show an example of the output that `imrestore` generates lets run the program like a normal user would do if he downloaded the repository from (David, 2016b) to use it just by placing in the console the command ‘`python imrestore.py tests/im1* --lens --overwrite --cachePath .`’.

Notice that the option `--cachePath .` is provided to create the cache in the current path of the terminal which generates a folder called *descriptors* with structure shown in Figure 36a. In the first try it executes everything normally but caching data for subsequent executions (see the output messages generated by `imrestore` inTable 3). In the second execution it finds that *descriptors* folder is already created so it looks for cached data when needed (see Table 4) to not process it and if data is missing in the cache it simply recalculates it. For the sake of demonstrating it lets manipulate the persisted data using *memoizedDict* provided by `RRtoolbox` package which is the developed cache manager used by `imrestore`. In a python environment the user only has to import the *memoizedDict* class and instantiate it to delete the desired key(s) as it is shown in Figure 36b. Figure 36c shows how the file structure in the *descriptors* folder has changed after its manipulation. If `imrestore` is run once again it encounters that there are no features for ‘`tests/im1_1.jpg`’ so it calculates them and because it had cached ‘`tests/im1_2.jpg`’ this data is not re-evaluated saving time and computer power (see the output messages in Table 5 to confirm this behaviour). Notice that, because customization is a common procedure when restoring images of a same set, `imrestore` provides the `--onlykeys` option to process the key-points of images without restoring them which only makes sense in combination with cache options and if the user wants to keep using images of the same set many times.

Figure 36. Persistence to *descriptors* folder and its manipulation



Now that we know the *imrestore* basic use we proceed to run the generic command for 27 sets that contain retinal images with severe noise cases hoping that *imrestore* can give us the best automated results. Of course each set presents different challenges containing images with different perspectives, illumination, flares, unfocused blurry areas and others from where *imrestore* has to automatically choose the best images of the set with the first as the *base image* for merging and stitching procedures. A pattern is not fed to *imrestore*, in fact the idea is to test it with any possible scenario and let it take automated decisions. After this, we analyse all the results and re-run the program with fitting options to customise how *imrestore* process some sets that could offer better results.

The general command to process the results is in Code 5. The `--lens` flag is used to simulate the retinal camera lens at the end of the restoration process, this with the intention of eliminating the blurry areas around the retinal area which is a common issue when images are obtained with mobile devices. The `--overwrite` option replaces images with the same name when saving the restoration result and `--cachePath {temp}` option is used to tell *imrestore* to cache data in the *RRtoolbox's temp* folder allowing the use of Code 5 in any path while saving the cache in the same place always. Naturally a custom path could also be provided as `--cachePath <my custom path>`.

Only the images that were used are shown in the result figures, so that the first image is the *base image* (e.g. Figure 38a and Figure 39a) used for its restoration and the subsequent images except the last the ones used to improve it (e.g. Table 6 does not have more images and Table 7 has Figure 39b). The last image is the result for the set using the *base image* and it is what *imrestore* saves as output (e.g. Figure 38b and Figure 39c). In addition to it, a profiling which is like a report is presented for each set to let the user see the times of each process in the program. Notice that this differs to the output messages presented in Tables 3, 4 and 5 in that it is generated by the internal *imrestore* profiler that can be imported using `from RRtoolbox.lib.root import Profiler`.

Code 5: General command to process the retinal sets

```
python imrestore.py <path to folder with images> --lens --overwrite -c <cache file>
```

Table 3

Output messages for first execution of python imrestore.py tests/im1 --lens --overwrite --cachePath . command*

```

Configured for retinal restoration...
Cache path is in ./descriptors
No. images 2...
Computing features...

Features 1/2... Processing features for
  ↳ tests/im1_1.jpg...
affine sampling: 1 / 43
affine sampling: 2 / 43
affine sampling: 3 / 43
affine sampling: 4 / 43
affine sampling: 5 / 43
affine sampling: 6 / 43
affine sampling: 7 / 43
affine sampling: 8 / 43
affine sampling: 9 / 43
affine sampling: 10 / 43
affine sampling: 11 / 43
affine sampling: 12 / 43
affine sampling: 13 / 43
affine sampling: 14 / 43
affine sampling: 15 / 43
affine sampling: 16 / 43
affine sampling: 17 / 43
affine sampling: 18 / 43
affine sampling: 19 / 43
affine sampling: 20 / 43
affine sampling: 21 / 43
affine sampling: 22 / 43
affine sampling: 23 / 43
affine sampling: 24 / 43
affine sampling: 25 / 43
affine sampling: 26 / 43
affine sampling: 27 / 43
affine sampling: 28 / 43
affine sampling: 29 / 43
affine sampling: 30 / 43
affine sampling: 31 / 43
affine sampling: 32 / 43
affine sampling: 33 / 43
affine sampling: 34 / 43
affine sampling: 35 / 43
affine sampling: 36 / 43
affine sampling: 37 / 43
affine sampling: 38 / 43
affine sampling: 39 / 43
affine sampling: 40 / 43
affine sampling: 41 / 43
affine sampling: 42 / 43
affine sampling: 43 / 43

Computed feature time was 2.561066 seconds
baseImage is tests/im1_1.jpg
Configured to sort by best matches
Restoring ...
Matching ...
Matching overall time was 0.016191 seconds
Merging ...
This image has been merged: tests/im1_2.jpg...
Merging overall time was 0.856332 seconds
Matching ...
All images have been merged...
Matching overall time was 0.000033 seconds
Restoring overall time was 0.872662 seconds
Post-processing ...
Post-processing overall time was 0.234878 seconds
Saved: tests/_restored_im1_1.jpg

Features 2/2... Processing features for
  ↳ tests/im1_2.jpg...
affine sampling: 1 / 43
affine sampling: 2 / 43
affine sampling: 3 / 43

```

Table 4

Output messages for second execution of python imrestore.py tests/im1 --lens --overwrite --cachePath . command*

Configured for retinal restoration...	Matching ...
Cache path is in ./descriptors	Matching overall time was 0.015764 seconds
No. images 2...	Merging ...
Computing features...	This image has been merged: tests/im1_2.jpg...
Features 1/2... tests/im1_1.jpg is cached...	Merging overall time was 0.829140 seconds
Features 2/2... tests/im1_2.jpg is cached...	Matching ...
Computed feature time was 0.190396 seconds	All images have been merged...
baseImage is tests/im1_1.jpg	Matching overall time was 0.000030 seconds
Configured to sort by best matches	Restoring overall time was 0.845043 seconds
Restoring ...	Post-processing ...
	Post-processing overall time was 0.226974 seconds
	Saved: tests/_restored_im1_1.jpg

Table 5

Output messages for execution of python imrestore.py tests/im1 --lens --overwrite --cachePath . command with modified cache*

Configured for retinal restoration...	affine sampling: 28 / 43
Cache path is in ./descriptors	affine sampling: 29 / 43
No. images 2...	affine sampling: 30 / 43
Computing features...	affine sampling: 31 / 43
Features 1/2... Processing features for → tests/im1_1.jpg...	affine sampling: 32 / 43
affine sampling: 1 / 43	affine sampling: 33 / 43
affine sampling: 2 / 43	affine sampling: 34 / 43
affine sampling: 3 / 43	affine sampling: 35 / 43
affine sampling: 4 / 43	affine sampling: 36 / 43
affine sampling: 5 / 43	affine sampling: 37 / 43
affine sampling: 6 / 43	affine sampling: 38 / 43
affine sampling: 7 / 43	affine sampling: 39 / 43
affine sampling: 8 / 43	affine sampling: 40 / 43
affine sampling: 9 / 43	affine sampling: 41 / 43
affine sampling: 10 / 43	affine sampling: 42 / 43
affine sampling: 11 / 43	affine sampling: 43 / 43
affine sampling: 12 / 43	Features 2/2... tests/im1_2.jpg is cached...
affine sampling: 13 / 43	Computed feature time was 1.246475 seconds
affine sampling: 14 / 43	baseImage is tests/im1_1.jpg
affine sampling: 15 / 43	Configured to sort by best matches
affine sampling: 16 / 43	Restoring ...
affine sampling: 17 / 43	Matching ...
affine sampling: 18 / 43	Matching overall time was 0.016467 seconds
affine sampling: 19 / 43	Merging ...
affine sampling: 20 / 43	This image has been merged: tests/im1_2.jpg...
affine sampling: 21 / 43	Merging overall time was 0.856636 seconds
affine sampling: 22 / 43	Matching ...
affine sampling: 23 / 43	All images have been merged...
affine sampling: 24 / 43	Matching overall time was 0.000033 seconds
affine sampling: 25 / 43	Restoring overall time was 0.873242 seconds
affine sampling: 26 / 43	Post-processing ...
affine sampling: 27 / 43	Post-processing overall time was 0.232622 seconds
	Saved: tests/_restored_im1_1.jpg

Result for set 1

Figure 37. Images in set1

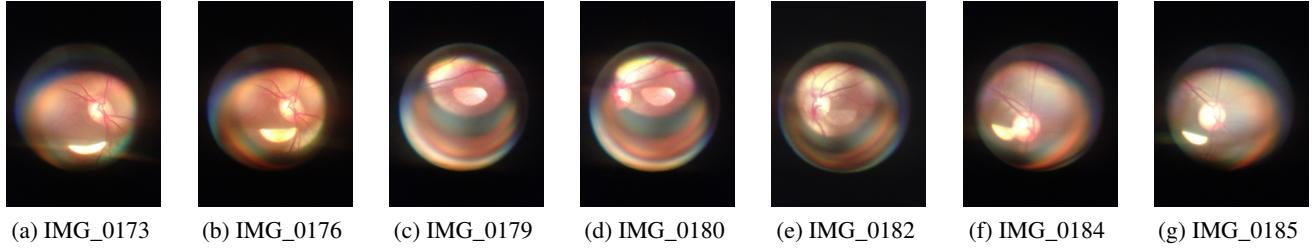


Table 6

Profiling for set1 using command ‘imrestore ../results/set1/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 11.4868 secs	'ImRestore init' -> 4.2917 secs
└── 'Computing features' -> 11.0730 secs	└── 'Computing features' -> 2.7749 secs
└── (processed) 'IMG_0173' -> 1.7930 secs	└── (memoized) 'IMG_0173' -> 0.6027 secs
└── (processed) 'IMG_0176' -> 1.5659 secs	└── (memoized) 'IMG_0176' -> 0.5151 secs
└── (processed) 'IMG_0179' -> 1.5224 secs	└── (memoized) 'IMG_0179' -> 0.3121 secs
└── (processed) 'IMG_0180' -> 1.5101 secs	└── (memoized) 'IMG_0180' -> 0.2485 secs
└── (processed) 'IMG_0182' -> 1.5386 secs	└── (memoized) 'IMG_0182' -> 0.2894 secs
└── (processed) 'IMG_0184' -> 1.5334 secs	└── (memoized) 'IMG_0184' -> 0.2105 secs
└── (processed) 'IMG_0185' -> 1.5580 secs	└── (memoized) 'IMG_0185' -> 0.4565 secs
└── 'Restoring' -> 0.1188 secs	└── 'Restoring' -> 0.3652 secs
└── 'Matching' -> 0.0535 secs	└── 'Matching' -> 0.1424 secs
└── 'Merging' -> 0.0650 secs	└── 'Merging' -> 0.2225 secs
└── 'Post-processing' -> 0.2951 secs	└── 'Post-processing' -> 1.1516 secs

Figure 38. Result for set1 using command ‘imrestore ../results/set1/*.* --lens --overwrite --cachePath {temp}’

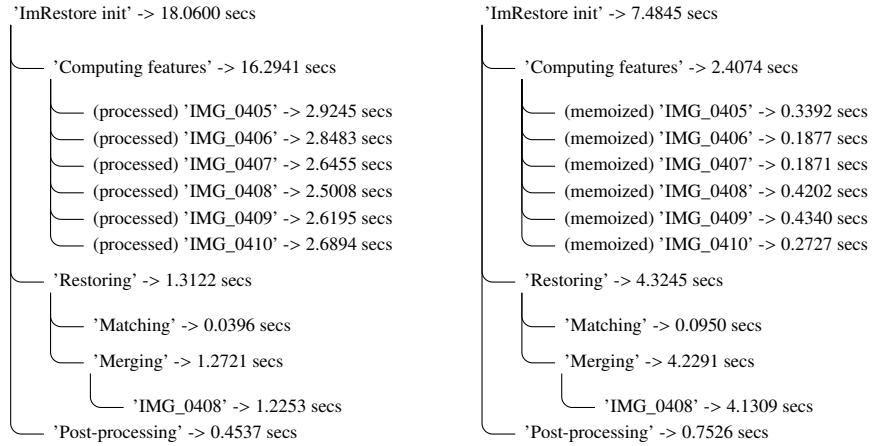


This set have 7 images as shown in Figure 37 and all of them are processed when Code 5 is used to find their features but only a few are selected by the algorithm in the restoration. This is evidenced in the entries ‘Computing features’ and ‘Restoring’ in Table 6 which shows the profiling of the execution. It demonstrates the advantage of using the cache option which reduces the un-cached time from 11.49 to 4.29 seconds when cached, this means that 7.20 seconds were spared with almost 3 times the gain.

Figure 38 shows the result for set 1. In it *imrestore* selects Figure 38a as the *base image* assuming that it is the clearest image with best features. Nonetheless this image has a flare that must be corrected but it does not find a good image from the set to achieve this so no additional images are used in the merging. Because of this Figure 38a is left unchanged and only post-processed to generate the restored image shown in Figure 38b.

Table 7

Profiling for set2 using command ‘imrestore ../results/set2/.* --lens --overwrite --cachePath {temp}’*



Result for set 2

Figure 39. Result for set2 using command ‘imrestore ../results/set2/.* --lens --overwrite --cachePath {temp}’*

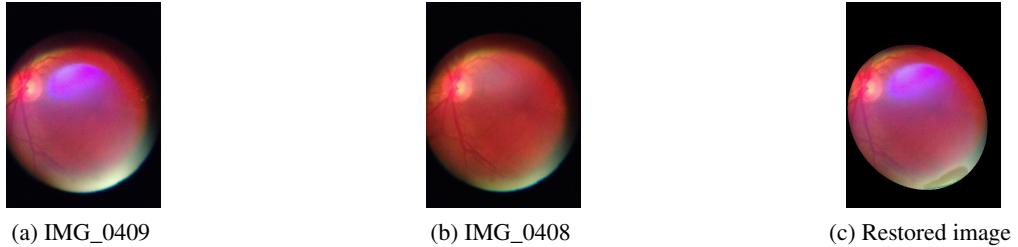
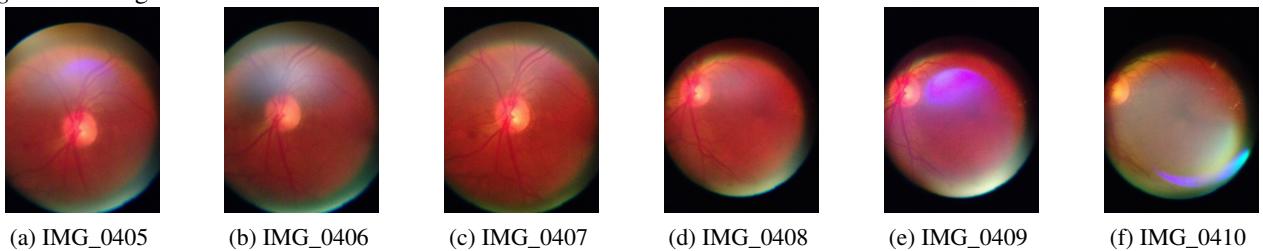


Figure 40. Images in set2



As in the Result for set 1 the generic Code 5 is used for set 2 with images in Figure 40 and presenting the profiling results in Table 7 for 6 processed images. It continues to show the advantage of using the cache option reducing the time from 18.06 to 7.48 seconds which is the 41.44% of the total with 10.58 seconds gained time.

Images used in the restoration are shown in Figure 39 and the ones not used in it are in ???. This time it does finds the merging image shown in Figure 39b to try to restore the *base image* shown in Figure 39a resulting in the restored image in Figure 39c. As it can be seen, to our judgement, the *base image* is not that appealing, in fact the *base image* could have been better the merged image in Figure 39b or the one which was not used in Figure 40c. This is due that in the form that *imrestore* is currently configured it assumes that the image with most key-points is the most likely to restore from.

Table 8

Profiling for set2 using command ‘imrestore ./results/set2/.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}’*

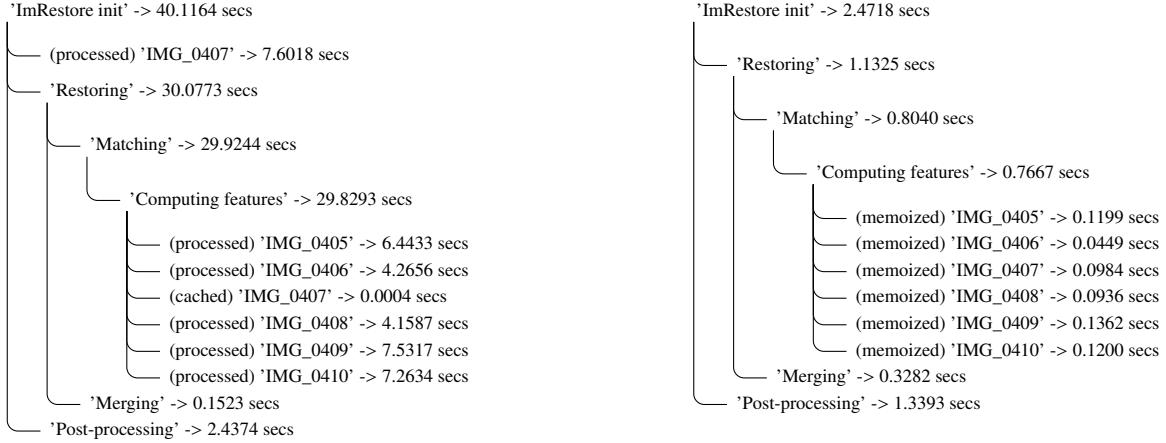


Figure 41. Customized result for set2 using command ‘imrestore ./results/set2/.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}’*



Lets try to enhance the result showed by Figure 39 created when using the generic command in Code 5. To do this I will introduce two more options, the `--denoise` and `--baseImage` flags. The flag `--denoise` is pretty obvious, it uses an automated algorithm to select the best parameters to apply filters to the image so that it can be smoother to the user. `--baseImage` also explains itself as it is used to select a custom *base image* or configure its automated selection. Now, lets select the new *base image* as `IMG_0407` using the pattern `'IMG_0407.*'` where the point and wildcat `*` is used to specify that it can find the image in the set with any extension. Notice that without `*.*` in the pattern the image cannot be found because there is no image in the set without extension leading to an error and also if there are more than one image with the same name but with different extensions the program would find more than one image with this pattern throwing an error indicating this. The customised command is then `'imrestore ./results/set2/*.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}'` which uses abbreviations of `--baseImage` and `--denoise` as `--b` and `--y` respectively. Figure 41 shows the new result, with this *imrestore* produced one of the best results that could be obtained from set2 and though it did not find good matches for Figure 41a it applied the post processing methods to generate the restored image in Figure 41b. In a way the algorithm did not find more matches to merge in the new *base image* due that it was sufficient and had no problems whatsoever for the other images in the set to enhance, these images were not that good (Figure 40).

Table 8 shows the profiling for both the processed and cached case of the custom command `'imrestore ./results/set2/*.* -b IMG_0407.* -y --lens --overwrite --cachePath {temp}'`. There shouldn't be nothing abnormal about it, as always the cached profiling register less time than the un-cached one reducing the time from 40.12 to 2.47 seconds, only the 6.16%. Wait! there is something different about this profiling than the presented in Table 7, the profiling for the session without `--cachePath` (left) processed `IMG_0407` features ahead of time, what could have happened? well, it is due to the capacity of the program to perform Lazy evaluations and because the user asked the program to use `IMG_0407` as the *base image* it had to process its features before the restoration process began. More intriguing is that this same image was requested when all the features were computed in the 'Computing features' entry (Table 8, left) but because it was all-ready processed the program returned the cached data. Though the option for caching was not specified the program by default is configured to use cache internally but not between sessions (recognized by the 'memoized' tag in the features when cached, Table 8 right), this behaviour can be changed using the `--clearCache` flag which offers more caching options. Why the cached session in the right of Table 8 did not register the same as its processed counter part? it was due that when the cache option is enabled (e.g. `--cachePath` is used) the program shares the information of previous sessions and no further processing of the quested data is needed, this can be dangerous

as previous data could change the normal results of the current session, this can be prevented using `--clearCache 1` option to check data integrity each time it is requested. These caching functionalities are provided by the LazyDict and MemoizedDict classes found in RRtoolbox/lib/cache.py and can be imported in Python with `from RRtoolbox.lib.cache import LazyDict, MemoizedDict`.

Result for set 3

Table 9

Profiling for set3 using command ‘imrestore ../results/set3/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 25.5017 secs	'ImRestore init' -> 9.4361 secs
'Computing features' -> 21.8672 secs	'Computing features' -> 2.1685 secs
(processed) 'IMG_0428' -> 3.5337 secs	(memoized) 'IMG_0428' -> 0.1825 secs
(processed) 'IMG_0429' -> 3.2418 secs	(memoized) 'IMG_0429' -> 0.3714 secs
(processed) 'IMG_0430' -> 2.9368 secs	(memoized) 'IMG_0430' -> 0.3066 secs
(processed) 'IMG_0431' -> 2.9219 secs	(memoized) 'IMG_0431' -> 0.3074 secs
(processed) 'IMG_0432' -> 2.9320 secs	(memoized) 'IMG_0432' -> 0.3854 secs
(processed) 'IMG_0433' -> 2.8682 secs	(memoized) 'IMG_0433' -> 0.1596 secs
(processed) 'IMG_0434' -> 3.0255 secs	(memoized) 'IMG_0434' -> 0.1585 secs
'Restoring' -> 3.1824 secs	'Restoring' -> 6.5249 secs
'Matching' -> 0.1331 secs	'Matching' -> 0.2359 secs
'Merging' -> 3.0486 secs	'Merging' -> 6.2884 secs
'IMG_0429' -> 1.4398 secs	'IMG_0429' -> 2.8025 secs
'IMG_0430' -> 1.5079 secs	'IMG_0430' -> 3.2572 secs
'Post-processing' -> 0.4521 secs	'Post-processing' -> 0.7427 secs

Figure 42. Result for set3 using command ‘imrestore ../results/set3/*.* --lens --overwrite --cachePath {temp}’

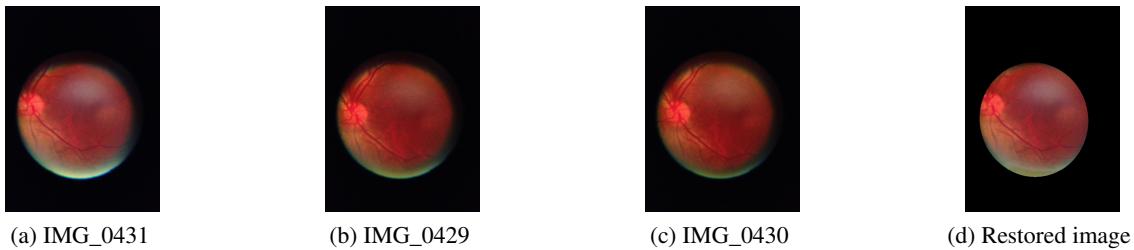
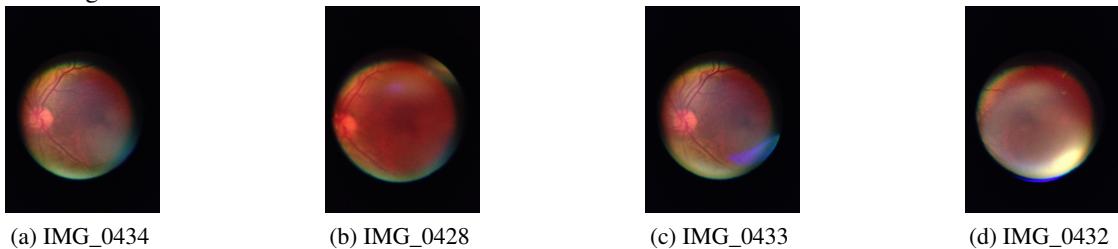


Figure 43. Images not used in restoration of set3



In Table 9 there are 7 computed images for their features and 2 in the ‘Merging’ entry. This is confirmed in Figure 42 with the *base image* in Figure 42a, the restored in Figure 42d and the remaining merged images. In the restored image we can see that the bright part at the bottom in Figure 42a was replaced by the one in Figure 42b which is more convenient. The lens also did a good job in segmenting correctly the retinal area despite the blurry-like aura enclosing the restoring image. The 3 used images were very similar and though the selected *base*

image was not the best of them this restoration can be considered good as the other images were not suitable presenting blurriness and violet lights with more whitish borders in the retinal area.

The explanation for the profiling tables are becoming superfluous as it seems that the cache will always reduce the processing time of an *imrestore* session compared to one that is not cached. In Table 9 is evidenced that it proudly reduced the time in a 63.00% from 25.50 to 9.44 seconds. The rate of used images also remains low in 0.43 with 3 used of 7 in the set (all images in Figure 42 except the restored in Figure 42d plus the 4 in Figure 43).

Result for set 4

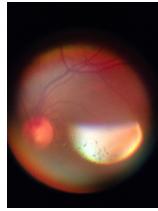
Table 10

Profiling for set4 using command ‘imrestore ..//results/set4/.* --lens --overwrite*

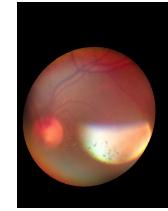
--cachePath {temp}

'ImRestore init' -> 37.1245 secs	'ImRestore init' -> 5.2054 secs
'Computing features' -> 36.5185 secs	'Computing features' -> 3.7392 secs
(processed) 'IMG_0411' -> 3.2775 secs	(memoized) 'IMG_0411' -> 0.3590 secs
(processed) 'IMG_0412' -> 3.1809 secs	(memoized) 'IMG_0412' -> 0.3026 secs
(processed) 'IMG_0413' -> 2.9239 secs	(memoized) 'IMG_0413' -> 0.1917 secs
(processed) 'IMG_0418' -> 2.9996 secs	(memoized) 'IMG_0418' -> 0.3780 secs
(processed) 'IMG_0419' -> 2.8301 secs	(memoized) 'IMG_0419' -> 0.1714 secs
(processed) 'IMG_0420' -> 2.8677 secs	(memoized) 'IMG_0420' -> 0.0978 secs
(processed) 'IMG_0421' -> 2.9634 secs	(memoized) 'IMG_0421' -> 0.0921 secs
(processed) 'IMG_0423' -> 2.8928 secs	(memoized) 'IMG_0423' -> 0.2538 secs
(processed) 'IMG_0424' -> 2.8513 secs	(memoized) 'IMG_0424' -> 0.3392 secs
(processed) 'IMG_0425' -> 2.9598 secs	(memoized) 'IMG_0425' -> 0.2417 secs
(processed) 'IMG_0426' -> 2.9678 secs	(memoized) 'IMG_0426' -> 0.8015 secs
(processed) 'IMG_0427' -> 3.0886 secs	(memoized) 'IMG_0427' -> 0.1747 secs
'Restoring' -> 0.1900 secs	'Restoring' -> 0.4057 secs
'Matching' -> 0.0540 secs	'Matching' -> 0.1222 secs
'Merging' -> 0.1356 secs	'Merging' -> 0.2830 secs
'Post-processing' -> 0.4161 secs	'Post-processing' -> 1.0605 secs

Figure 44. Result for set4 using command ‘imrestore ..//results/set4/*.* --lens --overwrite --cachePath {temp}’



(a) IMG_0411



(b) Restored image

This set presents similar results as the previous sets with 12 computed images and all of them with different affected areas as shown in Figure 44. Despite all those images only the image in <function <lambda> at 0x7f086072b320> got to be used without being merged with the others and only applying the post-processing step with lens simulation as shown in Figure 44b. There really is not a preferable image as all of them are not good and no image have desired parts that could be used to restore the *base image* but at least the enhancing of it demonstrated the capabilities of *imrestore* in situations were there is nothing to do.

Table 10 shows again a reduction in time of 31.92 seconds from 37.12 to 5.21 when using cache. Despite that the rate of used images is low with 0.08 as 0 images where merged leaving 1 used and 11 not used in the set of 12 images (Figure 45).

Figure 45. Images in set4

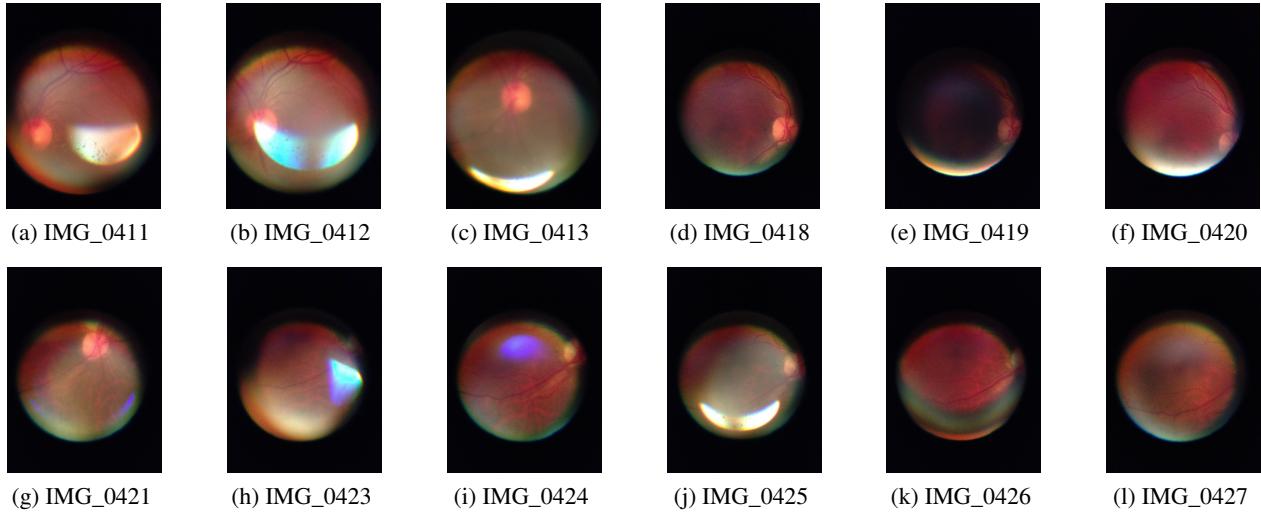


Table 11

```
Profiling for set5 using command 'imrestore ../results/set5/*.* --lens --overwrite
--cachePath {temp}'
```

'ImRestore init' -> 23.8650 secs	'ImRestore init' -> 5.9907 secs
└── 'Computing features' -> 21.7444 secs	└── 'Computing features' -> 3.4419 secs
└── (processed) 'IMG_0249' -> 3.2839 secs	└── (memoized) 'IMG_0249' -> 0.4656 secs
└── (processed) 'IMG_0250' -> 3.1891 secs	└── (memoized) 'IMG_0250' -> 0.6815 secs
└── (processed) 'IMG_0251' -> 3.0490 secs	└── (memoized) 'IMG_0251' -> 0.4374 secs
└── (processed) 'IMG_0252' -> 2.8878 secs	└── (memoized) 'IMG_0252' -> 0.2954 secs
└── (processed) 'IMG_0253' -> 2.9675 secs	└── (memoized) 'IMG_0253' -> 0.8914 secs
└── (processed) 'IMG_0254' -> 3.0005 secs	└── (memoized) 'IMG_0254' -> 0.2775 secs
└── (processed) 'IMG_0256' -> 2.9874 secs	└── (memoized) 'IMG_0256' -> 0.1578 secs
└── 'Restoring' -> 1.6591 secs	└── 'Restoring' -> 2.0090 secs
└── 'Matching' -> 0.0738 secs	└── 'Matching' -> 0.0854 secs
└── 'Merging' -> 1.5848 secs	└── 'Merging' -> 1.9230 secs
└── 'IMG_0254' -> 1.5285 secs	└── 'IMG_0254' -> 1.8020 secs
└── 'Post-processing' -> 0.4615 secs	└── 'Post-processing' -> 0.5398 secs

Result for set 5

In this set 2 images were used from 7 images leading to a rate of 0.29 which is evidenced in Figure 46. The *base image* is in Figure 46a an the other in Figure 46b. Both of them presented the same flare in the same position and though they could be perfectly matched they produced a similar flare which can be seen in Figure 46c. This led to the *base image* to not being adequately corrected due that all the other images presented the same problem but the lens simulation produced a more appealing result in the merged images. Notice that the images not used in the restoring process were blurrier and safely could be ignored (Figure 47).

For this set the profiling presented in Table 11 keeps generating satisfactory results of the *imrestore* performance. Even when the set have 7 images the restoration takes only 23.86 seconds in process their features, match all the images with the automatically selected *base image*, merge those which are convenient for the restoration and post-process them with a simulation of lens. But off course, this is when a basic configuration is applied, after all *imrestore* can be configured to apply filtering, histogram matching between merged images, change the behaviour of how the images are restored and more will be included in the future (see Future Work section). Even better the uncached time is bested when caching techniques are applied saving 17.87 seconds with its astonishing processing time of 5.99 seconds. The profiling puts in evidence that computing the feature is what takes most of the time in the application.

Figure 46. Result for set5 using command ‘imrestore ../results/set5/*.* --lens --overwrite --cachePath {temp}’

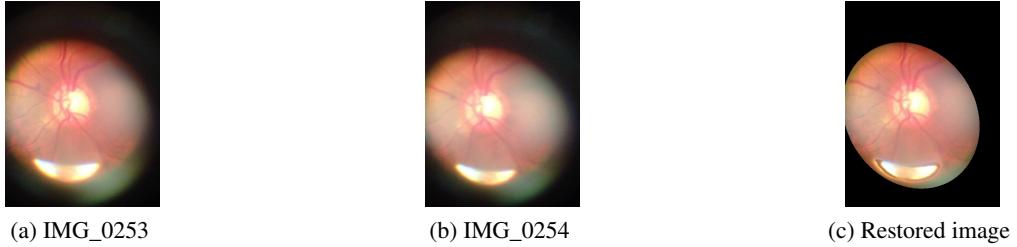
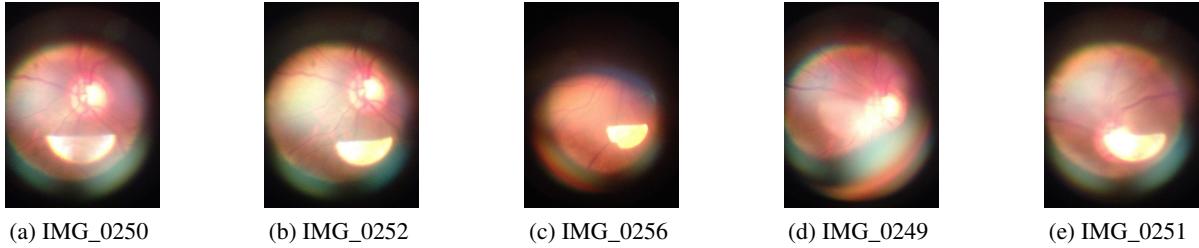


Figure 47. Images not used in restoration of set5



Result for set 6

Table 12

Profiling for set6 using command ‘imrestore ../results/set6/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 22.2521 secs	'ImRestore init' -> 3.1205 secs
└── 'Computing features' -> 21.6767 secs	└── 'Computing features' -> 1.8374 secs
└── (processed) 'IMG_0257' -> 3.2036 secs	└── (memoized) 'IMG_0257' -> 0.1445 secs
└── (processed) 'IMG_0258' -> 3.1040 secs	└── (memoized) 'IMG_0258' -> 0.1330 secs
└── (processed) 'IMG_0259' -> 2.9702 secs	└── (memoized) 'IMG_0259' -> 0.1316 secs
└── (processed) 'IMG_0260' -> 2.9582 secs	└── (memoized) 'IMG_0260' -> 0.2068 secs
└── (processed) 'IMG_0262' -> 2.9436 secs	└── (memoized) 'IMG_0262' -> 0.3962 secs
└── (processed) 'IMG_0263' -> 3.0434 secs	└── (memoized) 'IMG_0263' -> 0.2603 secs
└── (processed) 'IMG_0264' -> 3.1115 secs	└── (memoized) 'IMG_0264' -> 0.3681 secs
└── 'Restoring' -> 0.0625 secs	└── 'Restoring' -> 0.3728 secs
└── 'Matching' -> 0.0289 secs	└── 'Matching' -> 0.0326 secs
└── 'Merging' -> 0.0334 secs	└── 'Merging' -> 0.3399 secs
└── 'Post-processing' -> 0.5129 secs	└── 'Post-processing' -> 0.9103 secs

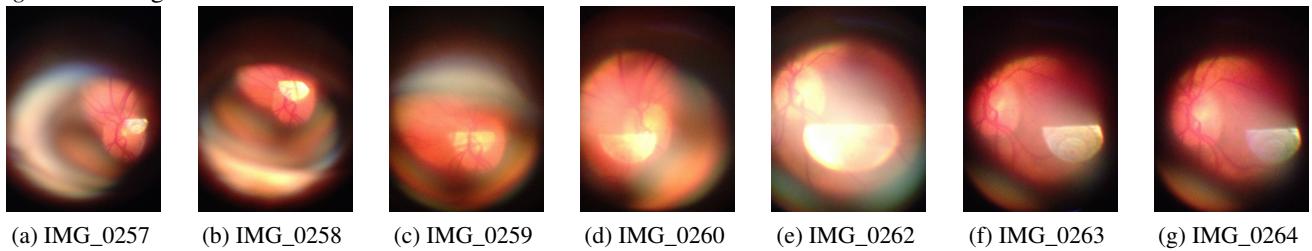
In Figure 48 there is an interesting case, from the 7 images in the set the one in Figure 48a got selected as the *base image* but it has some parts of the *Sclera* which is clearly a sign that this image should not be in the retinal set. Nonetheless the algorithm does not check for this and in fact selects the image proceeding as if nothing had happened. This case is similar to set5 in that all images are not good, none are suitable for restoring and most of them present parts of the eye that should not be in there.

As usual, the profiling of this set can be found in Table 12. It shows the 7 processed images in set6 shown in Figure 49. There was 0 images merged in the *base image* resulting in only one image used in total leaving a toll of 6 images not used for the set6. It is evidenced that the time was reduced by 7 times of the uncached session. In this way time passed from 22.25 to 3.12 seconds saving in total 19.13 seconds. This express a gain of the 85.98% with cached time only being 14.02%, this means that an excess in time of the 713.09% would happen if passed from cached to a not cached session. The statistics are not that encouraging with respect to the used images giving a rate of 0.14, 0.00 for merged images and 0.86 for failed images, a high rate for not used images.

Figure 48. Result for set6 using command ‘imrestore .../results/set6/*.* --lens --overwrite --cachePath {temp}’



Figure 49. Images in set6



In this set it is confirmed that the algorithm currently is not good dealing with parts that do not correspond to the retinal area nonetheless it applies the restoration method yielding still better results than the input images even if they are not satisfactory.

Result for set 7

Table 13

Profiling for set7 using command ‘imrestore .../results/set7/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 17.3964 secs	'ImRestore init' -> 6.2594 secs
└── 'Computing features' -> 15.7344 secs	└── 'Computing features' -> 1.8640 secs
└── (processed) 'IMG_0267' -> 3.4596 secs	└── (memoized) 'IMG_0267' -> 0.2475 secs
└── (processed) 'IMG_0269' -> 3.1190 secs	└── (memoized) 'IMG_0269' -> 0.3427 secs
└── (processed) 'IMG_0270' -> 3.1111 secs	└── (memoized) 'IMG_0270' -> 0.3933 secs
└── (processed) 'IMG_0271' -> 2.9327 secs	└── (memoized) 'IMG_0271' -> 0.4071 secs
└── (processed) 'IMG_0273' -> 3.0540 secs	└── (memoized) 'IMG_0273' -> 0.3494 secs
└── 'Restoring' -> 1.3234 secs	└── 'Restoring' -> 3.0931 secs
└── 'Matching' -> 0.0821 secs	└── 'Matching' -> 0.2248 secs
└── 'Merging' -> 1.2408 secs	└── 'Merging' -> 2.8679 secs
└── 'IMG_0271' -> 1.2214 secs	└── 'IMG_0271' -> 2.7820 secs
└── 'Post-processing' -> 0.3386 secs	└── 'Post-processing' -> 1.3023 secs

Using again Code 5 in this set *imrestore* processed 2 of the 5 images in it. The results can be appreciated in Figure 50 and the images not used in the restoration in Figure 51. There was a total of 1 merged images with the *base image* in Figure 50a and 3 unused images for the set7. The restored image in Figure 50c does not evidence any change from the initial image but the lens simulation did a good job. There was nothing that the merged image in Figure 50b could provide to it an so nothing was introduced. The other images where not selected obviously for their lack of information and affected areas that could worsen the restoring process of the set.

Table 13 presents the profiling produced for set7. It is evidenced that the time was reduced by 3 times of the uncached session. In this way time passed from 17.40 to 6.26 seconds saving in total 11.14 seconds. This is translated in a gain of the 64.02% with cache time being

Figure 50. Result for set7 using command ‘imrestore ../results/set7/*.* --lens --overwrite --cachePath {temp}’

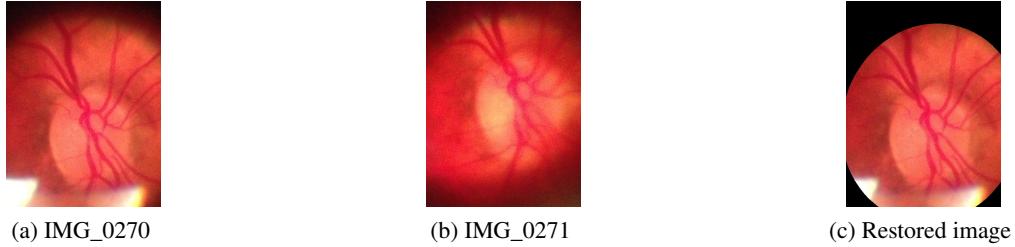
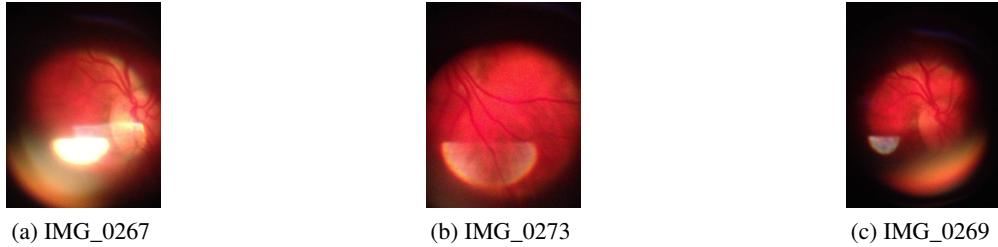


Figure 51. Images not used in restoration of set7



35.98% and this means that an excess in time of the 277.92% would happen if passed from cached to a not cached session. The rate of used images is 0.40, merged images 0.20 and failed images 0.60.

Result for set 8

Table 14

Profiling for set8 using command ‘imrestore ../results/set8/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 23.3291 secs	'ImRestore init' -> 4.5408 secs
'Computing features' -> 22.8403 secs	'Computing features' -> 2.7663 secs
(processed) 'IMG_0306' -> 3.5407 secs	(memoized) 'IMG_0306' -> 0.4620 secs
(processed) 'IMG_0307' -> 3.3176 secs	(memoized) 'IMG_0307' -> 0.2225 secs
(processed) 'IMG_0308' -> 3.1198 secs	(memoized) 'IMG_0308' -> 0.1624 secs
(processed) 'IMG_0309' -> 3.0648 secs	(memoized) 'IMG_0309' -> 0.2730 secs
(processed) 'IMG_0310' -> 3.1863 secs	(memoized) 'IMG_0310' -> 0.4763 secs
(processed) 'IMG_0311' -> 3.2326 secs	(memoized) 'IMG_0311' -> 0.2752 secs
(processed) 'IMG_0357' -> 3.0689 secs	(memoized) 'IMG_0357' -> 0.2454 secs
'Restoring' -> 0.0757 secs	'Restoring' -> 0.3333 secs
'Matching' -> 0.0285 secs	'Matching' -> 0.0767 secs
'Merging' -> 0.0470 secs	'Merging' -> 0.2561 secs
'Post-processing' -> 0.4130 secs	'Post-processing' -> 1.4412 secs

Figure 53 shows all the images presented in set8.

Figure 52 presents the results for set8 processing only 1 of the 7 images in it where the image in Figure 52a was selected as the *base image* leaving 6 images not used in the restoration of the set, this is shown in ???. The restored image in Figure 52b shows a huge improvement over the input image. It can be observed that though most of it was blurry around the retinal area it got segmented correctly and the lens simulation fitted perfectly. This would have not been achieved with normal shareholding methods that evidently would have added all the area around the retinal area.

Table 14 presents the profiling produced for set8 where we can observe the times which took the program to run each step of the restoration process. In it the uncached session time passed from 23.33 seconds to 4.54 seconds in the cached session saving in total 18.79

Figure 52. Result for set8 using command ‘imrestore .../results/set8/*.* --lens --overwrite --cachePath {temp}’

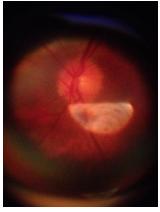


(a) IMG_0306

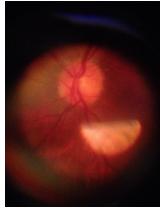


(b) Restored image

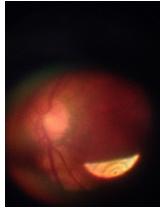
Figure 53. Images in set8



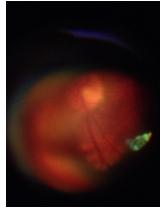
(a) IMG_0306



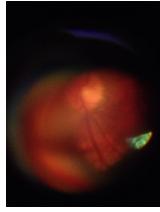
(b) IMG_0307



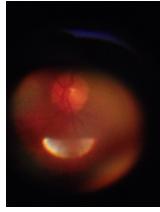
(c) IMG_0308



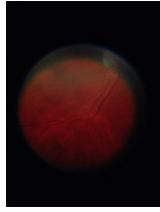
(d) IMG_0309



(e) IMG_0310



(f) IMG_0311



(g) IMG_0357

seconds that is 5 times faster than the uncached session and cache time only being the 19.46% from the total un-cached time.

The rate of used images is 0.14 with respect to the total of images in the set, this signifies that most images were not used but at the same time tells how well the algorithm left out the unwanted images. This is truth, as images in ?? present many problems and lack of information that cannot be used for the restoration of the *base image* in Figure 52a. The rate of images not used in the set is 0.86, completing the total with the used rate to 1. In a way these rates tell how good a set was taken for restoration, if it is low then images were not adequately taken but if the used rate reaches 1 or unused rate 0 it means that images present high quality from good perspectives of the same retinal image.

Result for set 9

Table 15

Profiling for set9 using command ‘imrestore ../results/set9/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 11.8629 secs	'ImRestore init' -> 7.0251 secs
'Computing features' -> 9.8378 secs	'Computing features' -> 1.6750 secs
(processed) 'IMG_0358' -> 3.4007 secs	(memoized) 'IMG_0358' -> 0.5123 secs
(processed) 'IMG_0359' -> 3.2114 secs	(memoized) 'IMG_0359' -> 0.7498 secs
(processed) 'IMG_0360' -> 3.0808 secs	(memoized) 'IMG_0360' -> 0.3352 secs
'Restoring' -> 1.5534 secs	'Restoring' -> 4.1770 secs
'Matching' -> 0.0342 secs	'Matching' -> 0.0747 secs
'Merging' -> 1.5187 secs	'Merging' -> 4.1017 secs
'IMG_0359' -> 1.5023 secs	'IMG_0359' -> 4.0529 secs
'Post-processing' -> 0.4718 secs	'Post-processing' -> 1.1731 secs

Figure 54. Result for set9 using command ‘imrestore ../results/set9/*.* --lens --overwrite --cachePath {temp}’

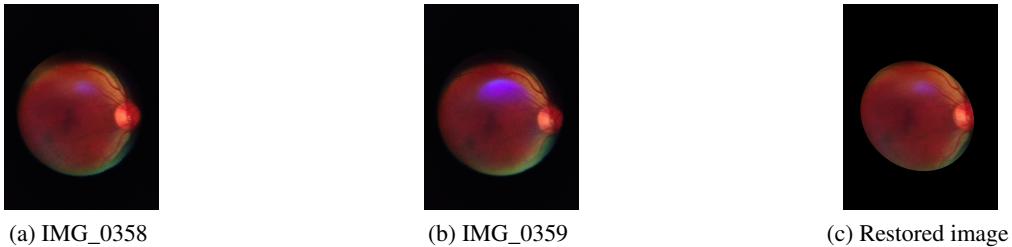
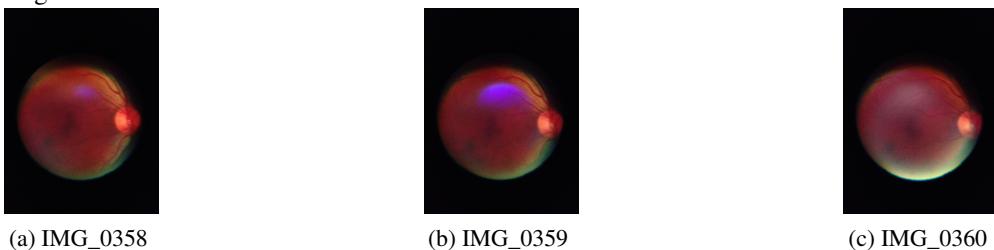


Figure 55. Images in set9



This set is only conformed by 3 images that are shown in Figure 55. Here no image present good quality and there is some kind of blue light in two of them (Figure 55a and Figure 55b), the other is too bright to be selected appreciated in Figure 55a. In spite of that 2 were used for the restoration with 1 used in Figure 54b to merge in the *base image* in Figure 54a and the other in Figure 55c was wisely left out from the restoration for its brightness. Figure 54 shows the merging process with the restored image in Figure 54c. The restored image is really similar to the *base image* with the difference of the lens simulation but it is ok as the merging image did not contribute good features to merge. In effect it was a good choice not to use any feature of the image in Figure 54b preventing the *base image* to be ruined.

The profiling produced for set9 is presented in Table 15 that shows prove of the restoration process and that it has not been altered. From it can extrated that the unchaed time was reduced 2 times with the cached session because the process took 11.86 seconds in the uncached session (lets assign it the 100%) and then 7.03 seconds in the cached session (the 59.22%), taking 4.84 seconds of difference (40.78%). If the cached session consumed the 100% of the processed time then the uncached time would be 168.87%.

This set performed a good used rate of 0.67 and of failed images as 0.33, indicating that in general the set was adequate for restoration. Notice however that these rates do not provide useful information regarding details like if the restoration was satisfactory or if the selected images did not present bad features.

Result for set 10

Table 16

Profiling for set10 using command ‘imrestore ../results/set10/.* --lens --overwrite --cachePath {temp}’*

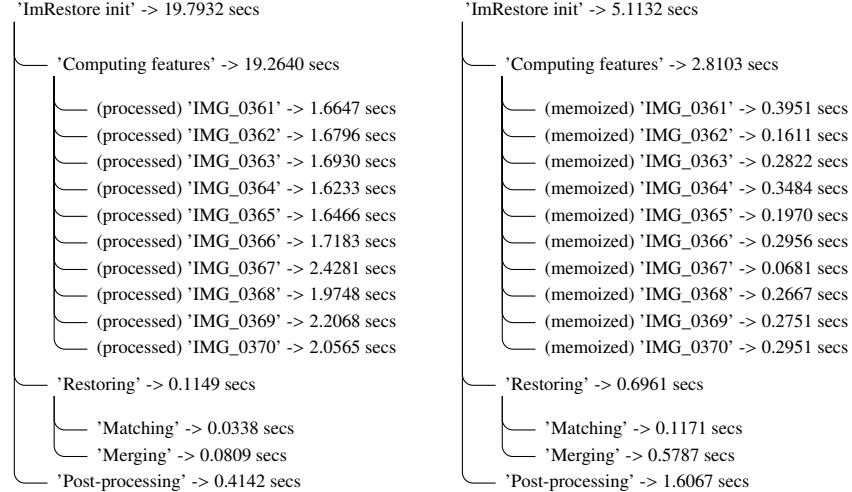
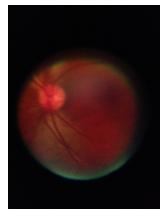


Figure 56. Result for set10 using command ‘imrestore ../results/set10/*.* --lens --overwrite --cachePath {temp}’

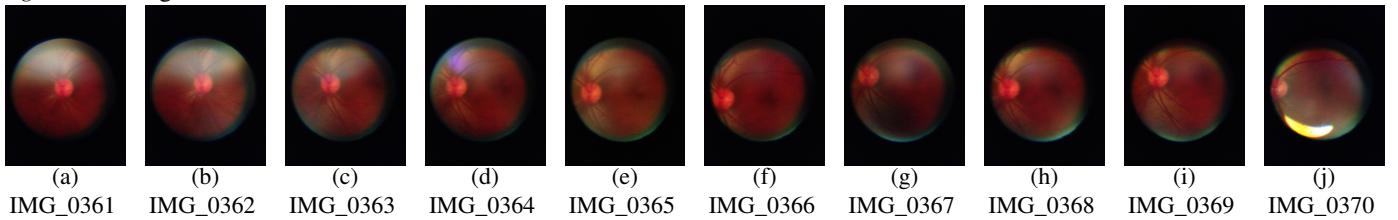


(a) IMG_0369



(b) Restored image

Figure 57. Images in set10



Using Code 5 in set10 *imrestore* processed 1 of the 10 images in set10, this is shown in Figure 57. 0 is the number of images merged in the *base image*. 9 is the number of images not used for the set10. Figure 56 presents the results for set10. ?? shows the images not used in set10. Figure 56a shows the *base image*, Figure 56b shows the restored image and the remaining merged image. Figure 57a shows the first image in set. ?? shows the first image not used in the merging. Table 16 presents the profiling produced for set10. It is evidenced that the time was reduced by 4 times of the uncached session. In this way time passed from 19.79 to 5.11 seconds saving in total 14.68 seconds. This is translated in a gain of the 74.17% with cache time being 25.83% and this means that an excess in time of the 387.10% would happen if passed from cached to a not cached session the rate of used images is 0.10. the rate of merged images is 0.00. the rate of failed images is 0.90.

Result for set 11

Table 17

Profiling for set11 using command ‘imrestore ../results/set11/.* --lens --overwrite --cachePath {temp}’*

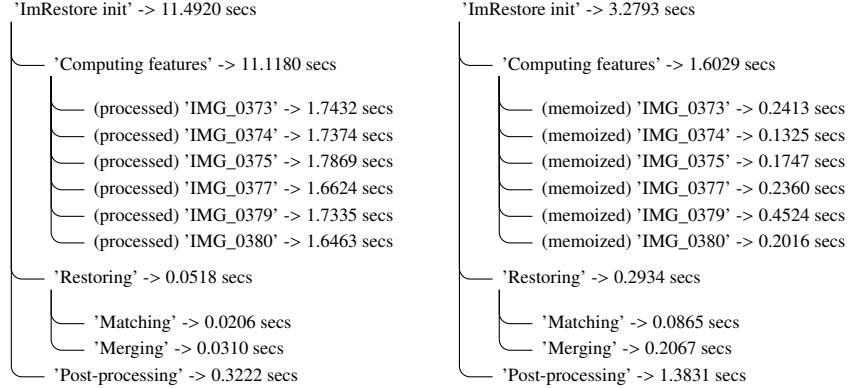
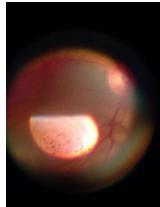
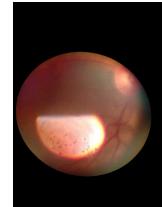


Figure 58. Result for set11 using command ‘imrestore ../results/set11/*.* --lens --overwrite --cachePath {temp}’



(a) IMG_0379

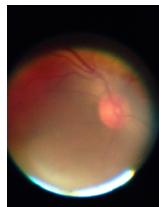


(b) Restored image

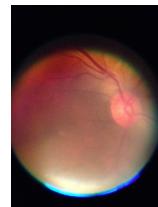
Figure 59. Images in set11



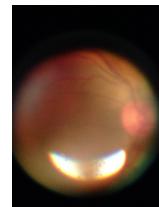
(a) IMG_0373



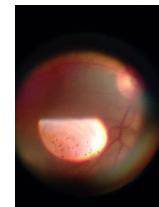
(b) IMG_0374



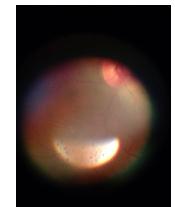
(c) IMG_0375



(d) IMG_0377



(e) IMG_0379



(f) IMG_0380

Using Code 5 in set11 *imrestore* processed 1 of the 6 images in set11, this is shown in Figure 59. 0 is the number of images merged in the *base image*. 5 is the number of images not used for the set11. Figure 58 presents the results for set11. ?? shows the images not used in set11. Figure 58a shows the *base image*, Figure 58b shows the restored image and the remaining merged image. Figure 59a shows the first image in set. ?? shows the first image not used in the merging. Table 17 presents the profiling produced for set11. It is evidenced that the time was reduced by 4 times of the uncached session. In this way time passed from 11.49 to 3.28 seconds saving in total 8.21 seconds. This is translated in a gain of the 71.46% with cache time being 28.54% and this means that an excess in time of the 350.44% would happen if passed from cached to a not cached session the rate of used images is 0.17. the rate of merged images is 0.00. the rate of failed images is 0.83.

Result for set 12

Table 18

Profiling for set12 using command ‘imrestore .. /results/set12/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 9.7331 secs	'ImRestore init' -> 3.4769 secs
└─ 'Computing features' -> 9.1343 secs	└─ 'Computing features' -> 1.9565 secs
└─ (processed) 'IMG_0381' -> 1.8524 secs	└─ (memoized) 'IMG_0381' -> 0.2367 secs
└─ (processed) 'IMG_0382' -> 2.0073 secs	└─ (memoized) 'IMG_0382' -> 0.5939 secs
└─ (processed) 'IMG_0383' -> 2.3685 secs	└─ (memoized) 'IMG_0383' -> 0.2115 secs
└─ (processed) 'IMG_0384' -> 2.7038 secs	└─ (memoized) 'IMG_0384' -> 0.3576 secs
└─ 'Restoring' -> 0.1217 secs	└─ 'Restoring' -> 0.2917 secs
└─ 'Matching' -> 0.0199 secs	└─ 'Matching' -> 0.0875 secs
└─ 'Merging' -> 0.1015 secs	└─ 'Merging' -> 0.1941 secs
└─ 'Post-processing' -> 0.4771 secs	└─ 'Post-processing' -> 1.2287 secs

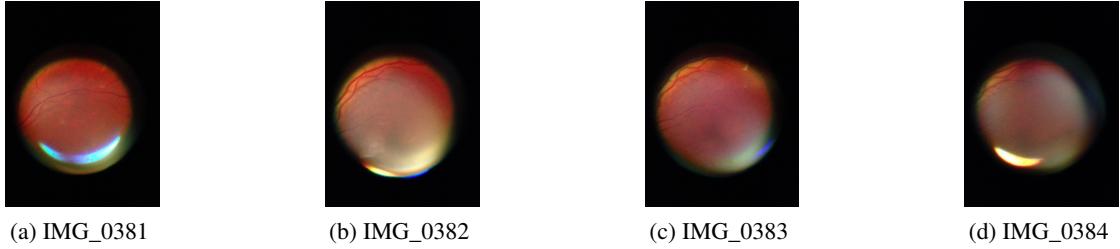
Figure 60. Result for set12 using command ‘imrestore .. /results/set12/*.* --lens --overwrite --cachePath {temp}’



(a) IMG_0382

(b) Restored image

Figure 61. Images in set12



(a) IMG_0381

(b) IMG_0382

(c) IMG_0383

(d) IMG_0384

Using Code 5 in set12 *imrestore* processed 1 of the 4 images in set12, this is shown in Figure 61. 0 is the number of images merged in the *base image*. 3 is the number of images not used for the set12. Figure 60 presents the results for set12. ?? shows the images not used in set12. Figure 60a shows the *base image*, Figure 60b shows the restored image and the remaining merged image. Figure 61a shows the first image in set. ?? shows the first image not used in the merging. Table 18 presents the profiling produced for set12. It is evidenced that the time was reduced by 3 times of the uncached session. In this way time passed from 9.73 to 3.48 seconds saving in total 6.26 seconds. This is translated in a gain of the 64.28% with cache time being 35.72% and this means that an excess in time of the 279.94% would happen if passed from cached to a not cached session the rate of used images is 0.25. the rate of merged images is 0.00. the rate of failed images is 0.75.

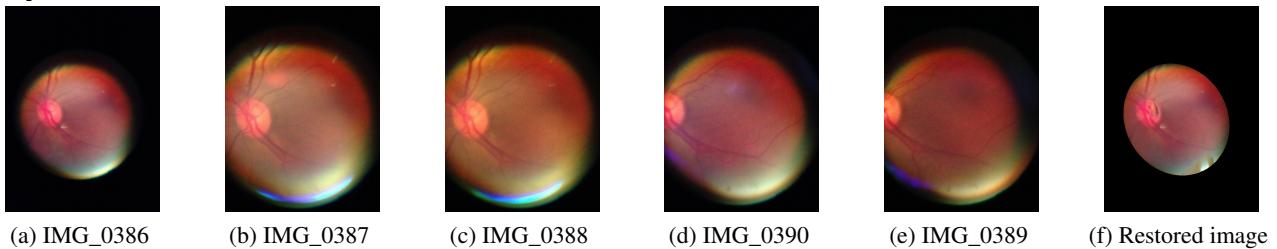
Result for set 13

Table 19

Profiling for set13 using command ‘imrestore/results/set13/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 18.3997 secs	'ImRestore init' -> 11.0310 secs
'Computing features' -> 12.8125 secs	'Computing features' -> 1.8227 secs
(processed) 'IMG_0386' -> 2.7297 secs	(memoized) 'IMG_0386' -> 0.3536 secs
(processed) 'IMG_0387' -> 2.6766 secs	(memoized) 'IMG_0387' -> 0.6477 secs
(processed) 'IMG_0388' -> 2.6601 secs	(memoized) 'IMG_0388' -> 0.1994 secs
(processed) 'IMG_0389' -> 2.1324 secs	(memoized) 'IMG_0389' -> 0.2435 secs
(processed) 'IMG_0390' -> 2.4107 secs	(memoized) 'IMG_0390' -> 0.2834 secs
'Restoring' -> 5.2833 secs	'Restoring' -> 8.5249 secs
'Matching' -> 0.0669 secs	'Matching' -> 0.0759 secs
'Merging' -> 5.2157 secs	'Merging' -> 8.4483 secs
'IMG_0387' -> 1.4573 secs	'IMG_0387' -> 1.8513 secs
'IMG_0388' -> 1.1390 secs	'IMG_0388' -> 2.0259 secs
'IMG_0390' -> 1.4722 secs	'IMG_0390' -> 1.9489 secs
'IMG_0389' -> 1.1470 secs	'IMG_0389' -> 2.6221 secs
'Post-processing' -> 0.3039 secs	'Post-processing' -> 0.6833 secs

Figure 62. Result for set13 using command ‘imrestore/results/set13/*.* --lens --overwrite --cachePath {temp}’



Using Code 5 in set13 *imrestore* processed 5 of the 5 images in set13, this is shown in ?? . 4 is the number of images merged in the *base image*. 0 is the number of images not used for the set13. Figure 62 presents the results for set13. ?? shows the images not used in set13. Figure 62a shows the *base image*, Figure 62f shows the restored image and the remaining merged images. ?? shows the first image in set. ?? shows the first image not used in the merging. Table 19 presents the profiling produced for set13. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 18.40 to 11.03 seconds saving in total 7.37 seconds. This is translated in a gain of the 40.05% with cache time being 59.95% and this means that an excess in time of the 166.80% would happen if passed from cached to a not cached session the rate of used images is 1.00. the rate of merged images is 0.80. the rate of failed images is 0.00.

Result for set 14

Table 20

Profiling for set14 using command ‘imrestore .../results/set14/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 15.8853 secs	'ImRestore init' -> 9.8135 secs
└─ 'Computing features' -> 12.4801 secs	└─ 'Computing features' -> 1.5526 secs
└─ (processed) 'IMG_0401' -> 2.7119 secs	└─ (memoized) 'IMG_0401' -> 0.2395 secs
└─ (processed) 'IMG_0402' -> 3.7434 secs	└─ (memoized) 'IMG_0402' -> 0.5507 secs
└─ (processed) 'IMG_0403' -> 3.1613 secs	└─ (memoized) 'IMG_0403' -> 0.3295 secs
└─ (processed) 'IMG_0404' -> 2.6146 secs	└─ (memoized) 'IMG_0404' -> 0.3061 secs
└─ 'Restoring' -> 2.9744 secs	└─ 'Restoring' -> 7.0761 secs
└─ 'Matching' -> 0.0961 secs	└─ 'Matching' -> 0.1532 secs
└─ 'Merging' -> 2.8776 secs	└─ 'Merging' -> 6.9223 secs
└─ 'IMG_0403' -> 1.4089 secs	└─ 'IMG_0403' -> 3.2824 secs
└─ 'IMG_0404' -> 1.4464 secs	└─ 'IMG_0404' -> 3.6093 secs
└─ 'Post-processing' -> 0.4309 secs	└─ 'Post-processing' -> 1.1848 secs

Figure 63. Result for set14 using command ‘imrestore .../results/set14/*.* --lens --overwrite --cachePath {temp}’

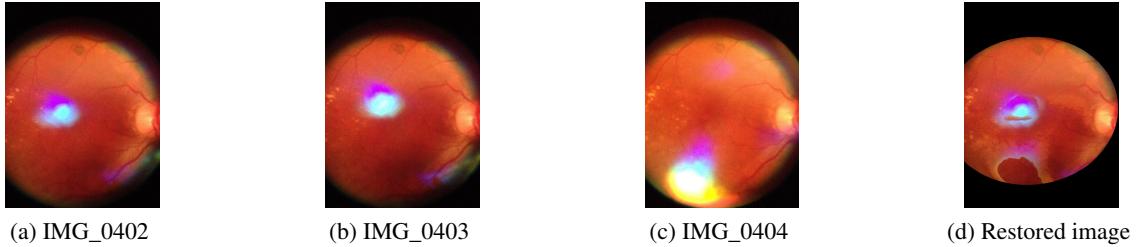
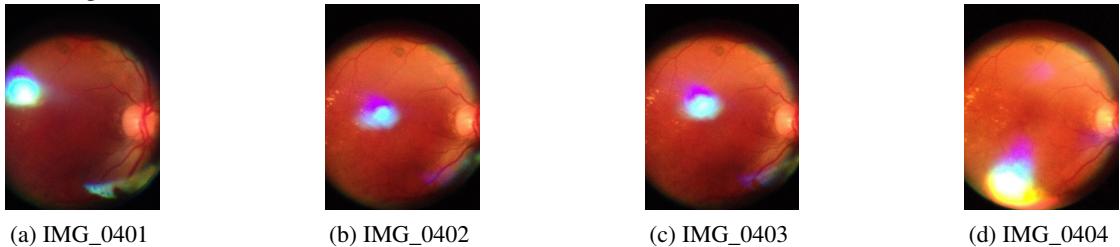


Figure 64. Images in set14



Using Code 5 in set14 *imrestore* processed 3 of the 4 images in set14, this is shown in Figure 64. 2 is the number of images merged in the *base image*. 1 is the number of images not used for the set14. Figure 63 presents the results for set14. ?? shows the images not used in set14. Figure 63a shows the *base image*, Figure 63d shows the restored image and the remaining merged images. Figure 64a shows the first image in set. ?? shows the first image not used in the merging. Table 20 presents the profiling produced for set14. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 15.89 to 9.81 seconds saving in total 6.07 seconds. This is translated in a gain of the 38.22% with cache time being 61.78% and this means that an excess in time of the 161.87% would happen if passed from cached to a not cached session the rate of used images is 0.75. the rate of merged images is 0.50. the rate of failed images is 0.25.

Result for set 15

Table 21

Profiling for set15 using command ‘imrestore .. /results/set15/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 64.7092 secs	'ImRestore init' -> 23.7269 secs
└─ 'Computing features' -> 56.7453 secs	└─ 'Computing features' -> 6.2923 secs
└─ (processed) '20150730_092749' -> 2.2607 secs	└─ (memoized) '20150730_092749' -> 0.2724 secs
└─ (processed) '20150730_092751' -> 2.0070 secs	└─ (memoized) '20150730_092751' -> 0.1949 secs
└─ (processed) '20150730_092752' -> 1.9977 secs	└─ (memoized) '20150730_092752' -> 0.1478 secs
└─ (processed) '20150730_092756' -> 2.0434 secs	└─ (memoized) '20150730_092756' -> 0.1412 secs
└─ (processed) '20150730_092757' -> 2.0984 secs	└─ (memoized) '20150730_092757' -> 0.1987 secs
└─ (processed) '20150730_092758(0)' -> 2.1579 secs	└─ (memoized) '20150730_092758(0)' -> 0.4324 secs
└─ (processed) '20150730_092758' -> 2.1397 secs	└─ (memoized) '20150730_092758' -> 0.3082 secs
└─ (processed) '20150730_092843' -> 2.2116 secs	└─ (memoized) '20150730_092843' -> 0.3421 secs
└─ (processed) '20150730_092846' -> 2.2935 secs	└─ (memoized) '20150730_092846' -> 0.6019 secs
└─ (processed) '20150730_092934' -> 2.2503 secs	└─ (memoized) '20150730_092934' -> 0.1055 secs
└─ (processed) '20150730_105818' -> 2.4197 secs	└─ (memoized) '20150730_105818' -> 0.1703 secs
└─ (processed) '20150730_105821' -> 2.4038 secs	└─ (memoized) '20150730_105821' -> 0.2121 secs
└─ (processed) '20150730_105822' -> 2.3793 secs	└─ (memoized) '20150730_105822' -> 0.1325 secs
└─ (processed) '20150730_105828' -> 2.5072 secs	└─ (memoized) '20150730_105828' -> 0.1290 secs
└─ (processed) '20150730_105830' -> 2.4340 secs	└─ (memoized) '20150730_105830' -> 0.1368 secs
└─ (processed) '20150730_105835' -> 2.5065 secs	└─ (memoized) '20150730_105835' -> 0.6456 secs
└─ (processed) '20150730_105842' -> 2.5571 secs	└─ (memoized) '20150730_105842' -> 0.3335 secs
└─ (processed) '20150730_105845' -> 2.6880 secs	└─ (memoized) '20150730_105845' -> 0.3172 secs
└─ (processed) '20150730_105901' -> 1.9463 secs	└─ (memoized) '20150730_105901' -> 0.1887 secs
└─ (processed) '20150730_105907' -> 1.8988 secs	└─ (memoized) '20150730_105907' -> 0.1632 secs
└─ (processed) '20150730_105912' -> 2.0264 secs	└─ (memoized) '20150730_105912' -> 0.3109 secs
└─ (processed) '20150730_105915' -> 3.2715 secs	└─ (memoized) '20150730_105915' -> 0.2279 secs
└─ (processed) '20150730_105921' -> 2.7538 secs	└─ (memoized) '20150730_105921' -> 0.1918 secs
└─ (processed) '20150730_105925' -> 2.6166 secs	└─ (memoized) '20150730_105925' -> 0.1654 secs
└─ 'Restoring' -> 7.5406 secs	└─ 'Restoring' -> 16.4187 secs
└─ 'Matching' -> 0.7458 secs	└─ 'Matching' -> 1.4788 secs
└─ 'Merging' -> 6.7934 secs	└─ 'Merging' -> 14.9385 secs
└─ '20150730_105835' -> 1.4151 secs	└─ '20150730_105835' -> 2.5812 secs
└─ '20150730_105818' -> 1.3202 secs	└─ '20150730_105818' -> 2.0717 secs
└─ '20150730_105821' -> 1.2991 secs	└─ '20150730_105821' -> 2.0427 secs
└─ '20150730_105828' -> 1.3112 secs	└─ '20150730_105828' -> 4.3992 secs
└─ '20150730_105822' -> 1.4006 secs	└─ '20150730_105822' -> 3.6504 secs
└─ 'Post-processing' -> 0.4233 secs	└─ 'Post-processing' -> 1.0160 secs

Using Code 5 in set15 *imrestore* processed 6 of the 24 images in set15, this is shown in ???. 5 is the number of images merged in the *base image*. 18 is the number of images not used for the set15. Figure 65 presents the results for set15. Figure 66 shows the images not used in set15. Figure 65a shows the *base image*, Figure 65g shows the restored image and the remaining merged images. ?? shows the first image in set. Figure 66a shows the first image not used in the merging. Table 21 presents the profiling produced for set15. It is evidenced that the time was reduced by 3 times of the uncached session. In this way time passed from 64.71 to 23.73 seconds saving in total 40.98 seconds. This is translated in a gain of the 63.33% with cache time being 36.67% and this means that an excess in time of the 272.72% would happen if passed from cached to a not cached session the rate of used images is 0.25. the rate of merged images is 0.21. the rate of failed images is 0.75.

Figure 65. Result for set15 using command ‘imrestore ./results/set15/*.* --lens --overwrite --cachePath {temp}’

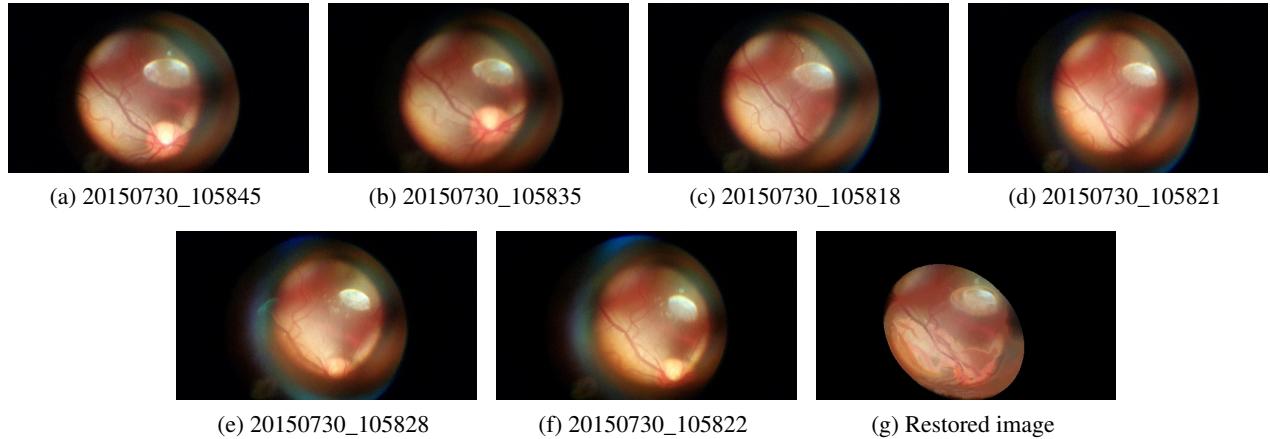
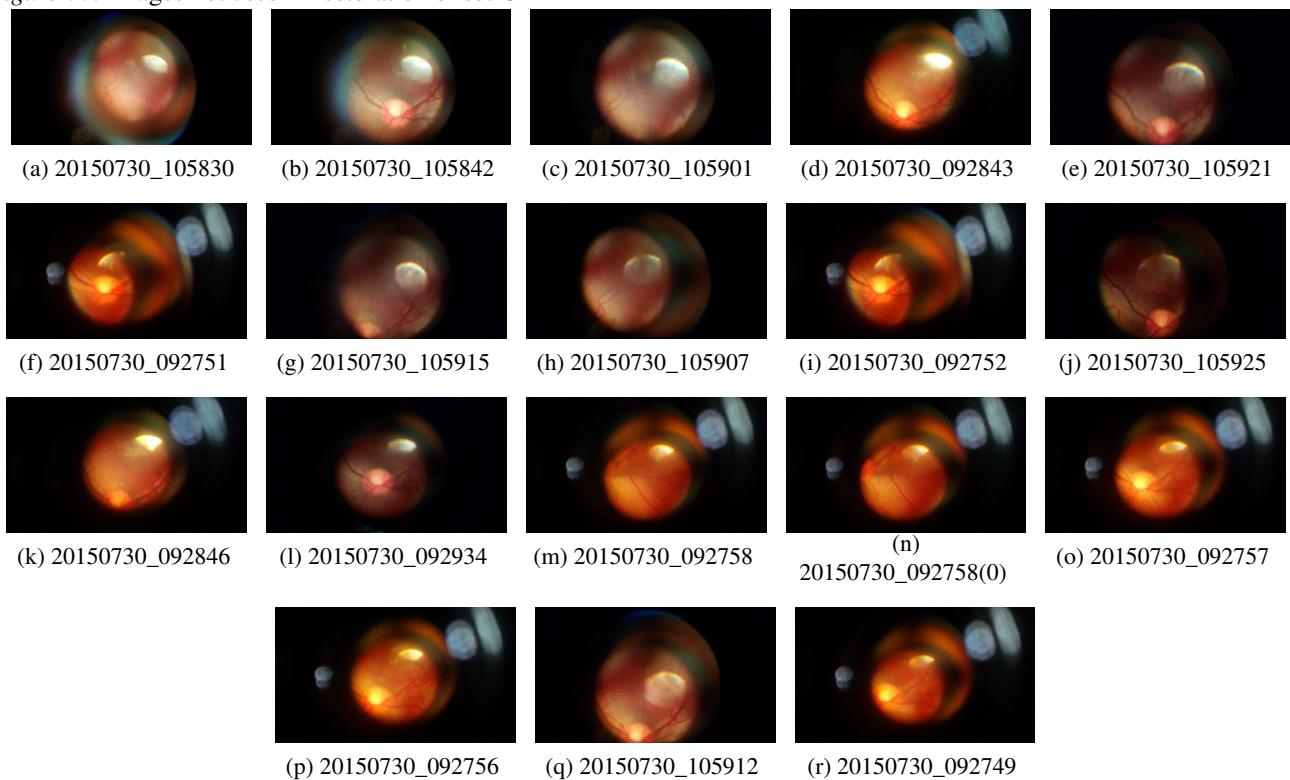
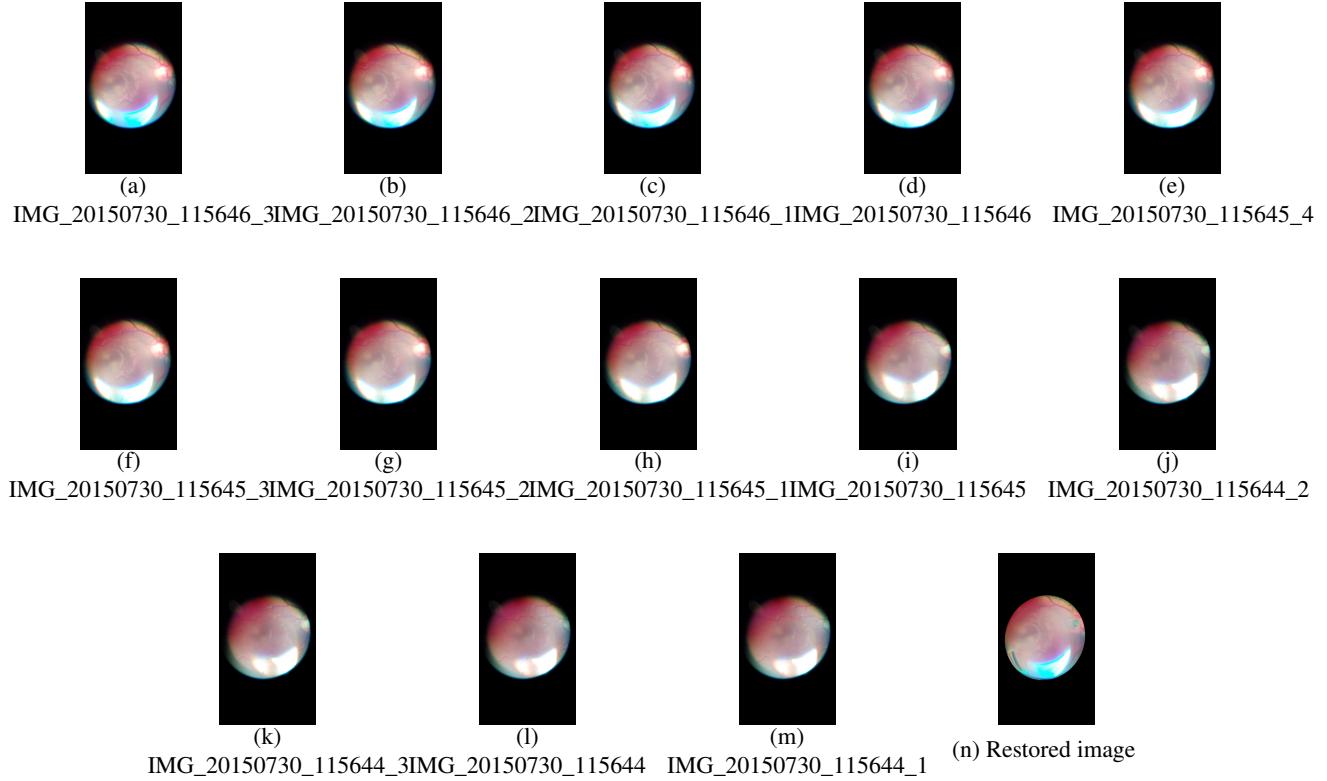


Figure 66. Images not used in restoration of set15



Result for set 16

Figure 67. Result for set16 using command ‘imrestore .. /results/set16/*.* --lens --overwrite --cachePath {temp}’,



Using Code 5 in set16 *imrestore* processed 13 of the 41 images in set16, this is shown in ???. 12 is the number of images merged in the *base image*. 28 is the number of images not used for the set16. Figure 67 presents the results for set16. ?? shows the images not used in set16. Figure 67a shows the *base image*, Figure 67n shows the restored image and the remaining merged images. ?? shows the first image in set. ?? shows the first image not used in the merging. Table 22 presents the profiling produced for set16. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 117.93 to 51.41 seconds saving in total 66.52 seconds. This is translated in a gain of the 56.41% with cache time being 43.59% and this means that an excess in time of the 229.40% would happen if passed from cached to a not cached session the rate of used images is 0.32. the rate of merged images is 0.29. the rate of failed images is 0.68.

Table 22

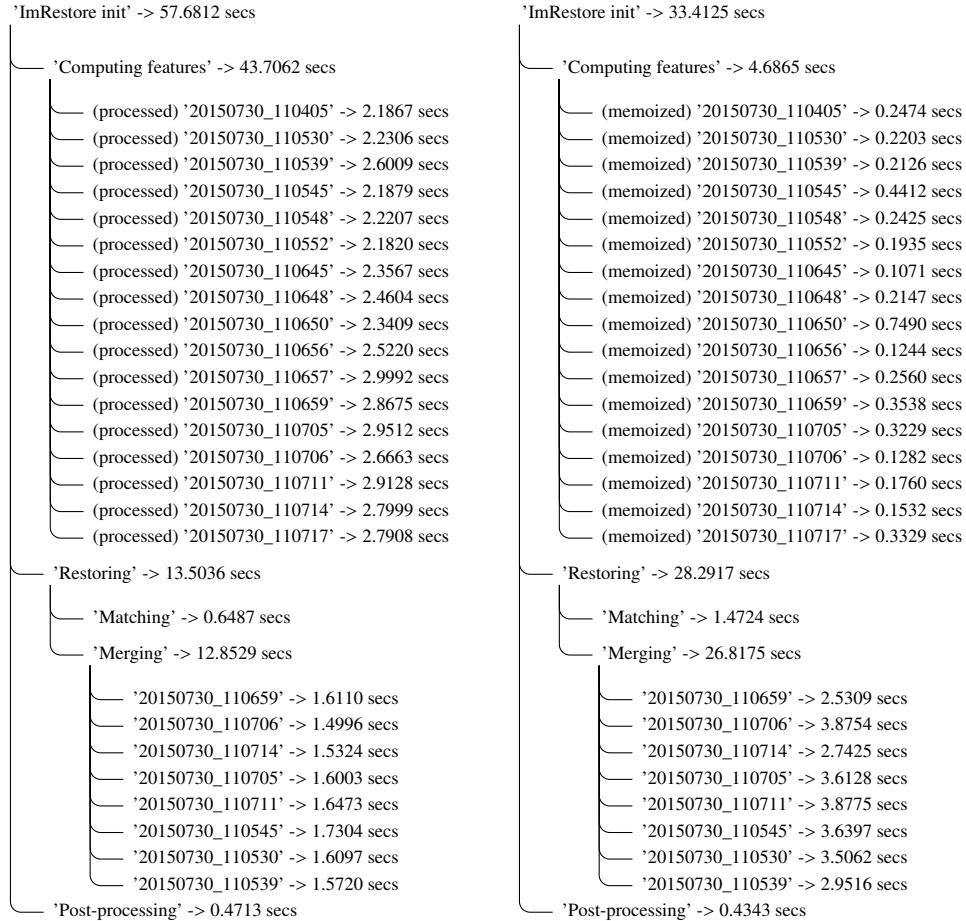
Profiling for set16 using command ‘imrestore ./results/set16/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 117.9345 secs	'ImRestore init' -> 51.4099 secs
'Computing features' -> 97.5388 secs	'Computing features' -> 2.9800 secs
(�processed) 'IMG_20150730_115646_2' -> 1.4922 secs	(�memoized) 'IMG_20150730_115646_2' -> 0.0569 secs
(�processed) 'IMG_20150730_115649_2' -> 1.4647 secs	(�memoized) 'IMG_20150730_115649_2' -> 0.0395 secs
(�processed) 'IMG_20150730_115642' -> 1.4554 secs	(�memoized) 'IMG_20150730_115642' -> 0.0501 secs
(�processed) 'IMG_20150730_115644' -> 1.4881 secs	(�memoized) 'IMG_20150730_115644' -> 0.0501 secs
(�processed) 'IMG_20150730_115644_1' -> 1.4671 secs	(�memoized) 'IMG_20150730_115644_1' -> 0.0381 secs
(�processed) 'IMG_20150730_115644_2' -> 1.4397 secs	(�memoized) 'IMG_20150730_115644_2' -> 0.0493 secs
(�processed) 'IMG_20150730_115644_3' -> 1.4666 secs	(�memoized) 'IMG_20150730_115644_3' -> 0.0400 secs
(�processed) 'IMG_20150730_115645' -> 1.5455 secs	(�memoized) 'IMG_20150730_115645' -> 0.0533 secs
(�processed) 'IMG_20150730_115645_1' -> 1.5194 secs	(�memoized) 'IMG_20150730_115645_1' -> 0.0420 secs
(�processed) 'IMG_20150730_115645_2' -> 1.5564 secs	(�memoized) 'IMG_20150730_115645_2' -> 0.0775 secs
(�processed) 'IMG_20150730_115645_3' -> 1.5483 secs	(�memoized) 'IMG_20150730_115645_3' -> 0.1020 secs
(�processed) 'IMG_20150730_115645_4' -> 1.5988 secs	(�memoized) 'IMG_20150730_115645_4' -> 0.1209 secs
(�processed) 'IMG_20150730_115646' -> 1.7309 secs	(�memoized) 'IMG_20150730_115646' -> 0.0653 secs
(�processed) 'IMG_20150730_115646_1' -> 2.4020 secs	(�memoized) 'IMG_20150730_115646_1' -> 0.0471 secs
(�processed) 'IMG_20150730_115646_3' -> 2.1936 secs	(�memoized) 'IMG_20150730_115646_3' -> 0.0686 secs
(�processed) 'IMG_20150730_115646_4' -> 2.1286 secs	(�memoized) 'IMG_20150730_115646_4' -> 0.0600 secs
(�processed) 'IMG_20150730_115646_5' -> 2.3441 secs	(�memoized) 'IMG_20150730_115646_5' -> 0.0434 secs
(�processed) 'IMG_20150730_115647' -> 2.5707 secs	(�memoized) 'IMG_20150730_115647' -> 0.0538 secs
(�processed) 'IMG_20150730_115647_1' -> 2.3986 secs	(�memoized) 'IMG_20150730_115647_1' -> 0.0777 secs
(�processed) 'IMG_20150730_115647_2' -> 2.3909 secs	(�memoized) 'IMG_20150730_115647_2' -> 0.0634 secs
(�processed) 'IMG_20150730_115647_3' -> 2.7939 secs	(�memoized) 'IMG_20150730_115647_3' -> 0.0406 secs
(�processed) 'IMG_20150730_115647_4' -> 2.4781 secs	(�memoized) 'IMG_20150730_115647_4' -> 0.0746 secs
(�processed) 'IMG_20150730_115647_5' -> 2.4089 secs	(�memoized) 'IMG_20150730_115647_5' -> 0.0536 secs
(�processed) 'IMG_20150730_115648' -> 2.7425 secs	(�memoized) 'IMG_20150730_115648' -> 0.0479 secs
(�processed) 'IMG_20150730_115648_1' -> 2.7123 secs	(�memoized) 'IMG_20150730_115648_1' -> 0.0669 secs
(�processed) 'IMG_20150730_115648_2' -> 2.9234 secs	(�memoized) 'IMG_20150730_115648_2' -> 0.0549 secs
(�processed) 'IMG_20150730_115648_3' -> 2.7251 secs	(�memoized) 'IMG_20150730_115648_3' -> 0.0473 secs
(�processed) 'IMG_20150730_115648_4' -> 2.8502 secs	(�memoized) 'IMG_20150730_115648_4' -> 0.0755 secs
(�processed) 'IMG_20150730_115649' -> 2.8096 secs	(�memoized) 'IMG_20150730_115649' -> 0.0457 secs
(�processed) 'IMG_20150730_115649_1' -> 2.6288 secs	(�memoized) 'IMG_20150730_115649_1' -> 0.0477 secs
(�processed) 'IMG_20150730_115649_3' -> 2.9014 secs	(�memoized) 'IMG_20150730_115649_3' -> 0.0533 secs
(�processed) 'IMG_20150730_115649_4' -> 3.0283 secs	(�memoized) 'IMG_20150730_115649_4' -> 0.0585 secs
(�processed) 'IMG_20150730_115649_5' -> 3.0345 secs	(�memoized) 'IMG_20150730_115649_5' -> 0.0535 secs
(�processed) 'IMG_20150730_115650' -> 3.0509 secs	(�memoized) 'IMG_20150730_115650' -> 0.0601 secs
(�processed) 'IMG_20150730_115650_1' -> 3.0572 secs	(�memoized) 'IMG_20150730_115650_1' -> 0.1502 secs
(�processed) 'IMG_20150730_115650_2' -> 3.3036 secs	(�memoized) 'IMG_20150730_115650_2' -> 0.0573 secs
(�processed) 'IMG_20150730_115650_3' -> 2.9822 secs	(�memoized) 'IMG_20150730_115650_3' -> 0.0566 secs
(�processed) 'IMG_20150730_115650_4' -> 3.0594 secs	(�memoized) 'IMG_20150730_115650_4' -> 0.0473 secs
(�processed) 'IMG_20150730_115651' -> 3.0236 secs	(�memoized) 'IMG_20150730_115651' -> 0.0609 secs
(�processed) 'IMG_20150730_115651_1' -> 2.9403 secs	(�memoized) 'IMG_20150730_115651_1' -> 0.0549 secs
(�processed) 'IMG_20150730_115651_2' -> 3.0323 secs	(�memoized) 'IMG_20150730_115651_2' -> 0.0663 secs
'Restoring' -> 19.9346 secs	'Restoring' -> 47.8004 secs
'Matching' -> 2.1919 secs	'Matching' -> 4.8422 secs
'Merging' -> 17.7395 secs	'Merging' -> 42.9549 secs
'IMG_20150730_115646_2' -> 1.2835 secs	'IMG_20150730_115646_2' -> 2.5186 secs
'IMG_20150730_115646_1' -> 1.3223 secs	'IMG_20150730_115646_1' -> 2.1790 secs
'IMG_20150730_115646' -> 1.3512 secs	'IMG_20150730_115646' -> 4.4143 secs
'IMG_20150730_115645_4' -> 1.4103 secs	'IMG_20150730_115645_4' -> 4.8933 secs
'IMG_20150730_115645_3' -> 1.4844 secs	'IMG_20150730_115645_3' -> 4.0004 secs
'IMG_20150730_115645_2' -> 1.4223 secs	'IMG_20150730_115645_2' -> 3.6895 secs
'IMG_20150730_115645_1' -> 1.4780 secs	'IMG_20150730_115645_1' -> 4.9356 secs
'IMG_20150730_115645' -> 1.3184 secs	'IMG_20150730_115645' -> 5.4561 secs
'IMG_20150730_115644_2' -> 1.4152 secs	'IMG_20150730_115644_2' -> 4.6674 secs
'IMG_20150730_115644_3' -> 1.4898 secs	'IMG_20150730_115644_3' -> 2.3625 secs
'IMG_20150730_115644' -> 1.5109 secs	'IMG_20150730_115644' -> 1.6333 secs
'IMG_20150730_115644_1' -> 1.6267 secs	'IMG_20150730_115644_1' -> 1.5454 secs
'Post-processing' -> 0.4611 secs	'Post-processing' -> 0.6295 secs

Result for set 17

Table 23

Profiling for set17 using command ‘imrestore .../results/set17/.* --lens --overwrite --cachePath {temp}’*



Using Code 5 in set17 *imrestore* processed 9 of the 17 images in set17, this is shown in ???. 8 is the number of images merged in the *base image*. 8 is the number of images not used for the set17. Figure 68 presents the results for set17. Figure 69 shows the images not used in set17. Figure 68a shows the *base image*, Figure 68j shows the restored image and the remaining merged images. ?? shows the first image in set. Figure 69a shows the first image not used in the merging. Table 23 presents the profiling produced for set17. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 57.68 to 33.41 seconds saving in total 24.27 seconds. This is translated in a gain of the 42.07% with cache time being 57.93% and this means that an excess in time of the 172.63% would happen if passed from cached to a not cached session the rate of used images is 0.53. the rate of merged images is 0.47. the rate of failed images is 0.47.

Figure 68. Result for set17 using command ‘imrestore ../results/set17/*.* --lens --overwrite --cachePath {temp}’

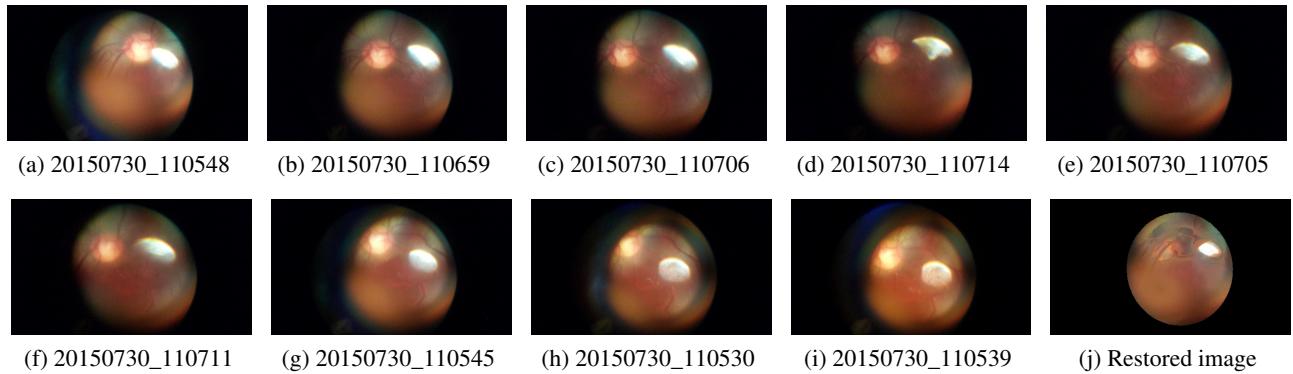
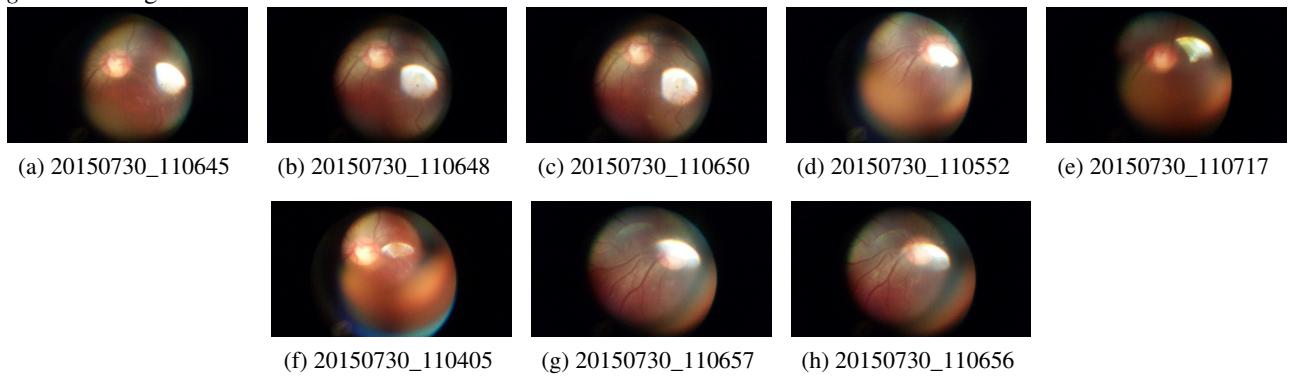


Figure 69. Images not used in restoration of set17



Result for set 18

Table 24

Profiling for set18 using command ‘imrestore/results/set18/.* --lens --overwrite --cachePath {temp}’*

```
'ImRestore init' -> 40.2444 secs
└── 'Computing features' -> 38.0787 secs
    ├── (processed) 'IMG_20150730_105626' -> 2.1840 secs
    ├── (processed) 'IMG_20150730_105632' -> 2.0535 secs
    ├── (processed) 'IMG_20150730_105655' -> 2.2025 secs
    ├── (processed) 'IMG_20150730_105748' -> 2.0853 secs
    ├── (processed) 'IMG_20150730_105753' -> 2.1469 secs
    ├── (processed) 'IMG_20150730_105800' -> 2.1074 secs
    ├── (processed) 'IMG_20150730_115251' -> 2.1841 secs
    ├── (processed) 'IMG_20150730_115254' -> 2.2393 secs
    ├── (processed) 'IMG_20150730_115301' -> 2.2283 secs
    ├── (processed) 'IMG_20150730_131444' -> 2.1268 secs
    ├── (processed) 'IMG_20150730_131454' -> 2.2785 secs
    ├── (processed) 'IMG_20150730_134219' -> 2.1742 secs
    ├── (processed) 'IMG_20150730_134225' -> 2.3129 secs
    ├── (processed) 'IMG_20150730_134231' -> 2.3135 secs
    ├── (processed) 'IMG_20150730_134235' -> 2.3683 secs
    ├── (processed) 'IMG_20150730_134254' -> 2.3516 secs
    └── (processed) 'IMG_20150730_134300' -> 2.4463 secs

└── 'Restoring' -> 1.7388 secs
    ├── 'Matching' -> 0.2569 secs
    ├── 'Merging' -> 1.4814 secs
        └── 'IMG_20150730_105748' -> 1.3090 secs
    'Post-processing' -> 0.4269 secs

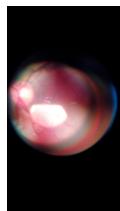
'ImRestore init' -> 11.1511 secs
└── 'Computing features' -> 6.1360 secs
    ├── (memoized) 'IMG_20150730_105626' -> 1.0331 secs
    ├── (memoized) 'IMG_20150730_105632' -> 0.2301 secs
    ├── (memoized) 'IMG_20150730_105655' -> 0.2322 secs
    ├── (memoized) 'IMG_20150730_105748' -> 0.1579 secs
    ├── (memoized) 'IMG_20150730_105753' -> 0.1915 secs
    ├── (memoized) 'IMG_20150730_105800' -> 0.3220 secs
    ├── (memoized) 'IMG_20150730_115251' -> 0.3197 secs
    ├── (memoized) 'IMG_20150730_115254' -> 0.5842 secs
    ├── (memoized) 'IMG_20150730_115301' -> 0.1764 secs
    ├── (memoized) 'IMG_20150730_131444' -> 0.1325 secs
    ├── (memoized) 'IMG_20150730_131454' -> 0.4424 secs
    ├── (memoized) 'IMG_20150730_134219' -> 0.2878 secs
    ├── (memoized) 'IMG_20150730_134225' -> 0.3203 secs
    ├── (memoized) 'IMG_20150730_134231' -> 0.1847 secs
    ├── (memoized) 'IMG_20150730_134235' -> 0.1742 secs
    ├── (memoized) 'IMG_20150730_134254' -> 0.1547 secs
    └── (memoized) 'IMG_20150730_134300' -> 0.1961 secs

└── 'Restoring' -> 4.2545 secs
    ├── 'Matching' -> 0.3605 secs
    ├── 'Merging' -> 3.8935 secs
        └── 'IMG_20150730_105748' -> 3.1016 secs
    'Post-processing' -> 0.7606 secs
```

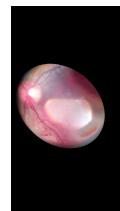
Figure 70. Result for set18 using command ‘imrestore ../../results/set18/*.* --lens --overwrite --cachePath {temp}’



(a) IMG_20150730_105626



(b) IMG_20150730_105748



(c) Restored image

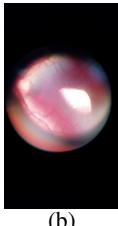
Using Code 5 in set18 *inrestore* processed 2 of the 17 images in set18, this is shown in Figure 71. 1 is the number of images merged in the *base image*. 15 is the number of images not used for the set18. Figure 70 presents the results for set18. ?? shows the images not used in set18. Figure 70a shows the *base image*, Figure 70c shows the restored image and the remaining merged images. Figure 71a shows the first image in set. ?? shows the first image not used in the merging. Table 24 presents the profiling produced for set18. It is evidenced that the time was reduced by 4 times of the uncached session. In this way time passed from 40.24 to 11.15 seconds saving in total 29.09 seconds. This is translated in a gain of the 72.29% with cache time being 27.71% and this means that an excess in time of the 360.90% would happen if passed from cached to a not cached session the rate of used images is 0.12. the rate of merged images is 0.06. the rate of failed images is 0.88.

Figure 71. Images in set18



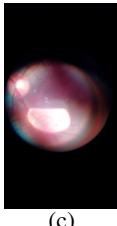
(a)

IMG_20150730_105626



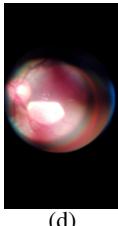
(b)

IMG_20150730_105632



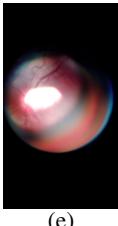
(c)

IMG_20150730_105655



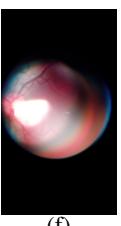
(d)

IMG_20150730_105748



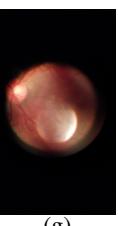
(e)

IMG_20150730_105753



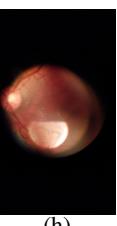
(f)

IMG_20150730_105800



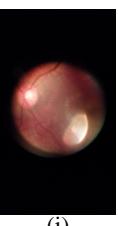
(g)

IMG_20150730_115251



(h)

IMG_20150730_115254



(i)

IMG_20150730_115301



(j)

IMG_20150730_131444



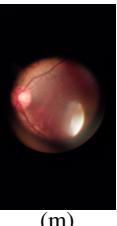
(k)

IMG_20150730_131454



(l)

IMG_20150730_134219



(m)

IMG_20150730_134225



(n)

IMG_20150730_134231



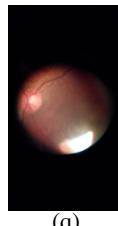
(o)

IMG_20150730_134235



(p)

IMG_20150730_134254



(q)

IMG_20150730_134300

Result for set 19

Using Code 5 in set18 *imrestore* processed 2 of the 17 images in set18, this is shown in Figure 71. 1 is the number of images merged in the *base image*. 15 is the number of images not used for the set18. Figure 70 presents the results for set18. ?? shows the images not used in set18. Figure 70a shows the *base image*, Figure 70c shows the restored image and the remaining merged images. Figure 71a shows the first image in set. ?? shows the first image not used in the merging. Table 24 presents the profiling produced for set18. It is evidenced that the time was reduced by 4 times of the uncached session. In this way time passed from 40.24 to 11.15 seconds saving in total 29.09 seconds. This is translated in a gain of the 72.29% with cache time being 27.71% and this means that an excess in time of the 360.90% would happen if passed from cached to a not cached session the rate of used images is 0.12. the rate of merged images is 0.06. the rate of failed images is 0.88.

```
In[9]: print generic_paragraph(s="set19",s2=None)
```

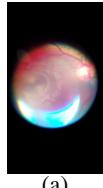
Using Code 5 in set19 *imrestore* processed 25 of the 31 images in set19, this is shown in ?? . 24 is the number of images merged in the *base image*. 6 is the number of images not used for the set19. Figure 72 presents the results for set19. Figure 73 shows the images not used in set19. Figure 72a shows the *base image*, Figure 72z shows the restored image and the remaining merged images. ?? shows the first image in set. Figure 73a shows the first image not used in the merging. Table 25 presents the profiling produced for set19. It is evidenced that the time was reduced by 1 times of the uncached session. In this way time passed from 120.48 to 92.78 seconds saving in total 27.70 seconds. This is translated in a gain of the 22.99% with cache time being 77.01% and this means that an excess in time of the 129.85% would happen if passed from cached to a not cached session the rate of used images is 0.81. the rate of merged images is 0.77. the rate of failed images is 0.19.

Table 25

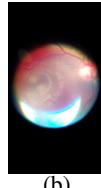
Profiling for set19 using command ‘imrestore .../results/set19/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 120.4802 secs	'ImRestore init' -> 92.7849 secs
-'Computing features' -> 86.0512 secs	-'Computing features' -> 8.0022 secs
-'(processed) 'IMG_20150730_115647_1' -> 2.8565 secs	-'(memoized) 'IMG_20150730_115647_1' -> 0.2936 secs
-'(processed) 'IMG_20150730_115642' -> 2.8028 secs	-'(memoized) 'IMG_20150730_115642' -> 0.2081 secs
-'(processed) 'IMG_20150730_115644' -> 2.6503 secs	-'(memoized) 'IMG_20150730_115644' -> 0.2033 secs
-'(processed) 'IMG_20150730_115644_1' -> 2.5347 secs	-'(memoized) 'IMG_20150730_115644_1' -> 0.2508 secs
-'(processed) 'IMG_20150730_115644_2' -> 2.6401 secs	-'(memoized) 'IMG_20150730_115644_2' -> 0.3342 secs
-'(processed) 'IMG_20150730_115644_3' -> 2.5682 secs	-'(memoized) 'IMG_20150730_115644_3' -> 0.3694 secs
-'(processed) 'IMG_20150730_115645_1' -> 2.6394 secs	-'(memoized) 'IMG_20150730_115645_1' -> 0.2020 secs
-'(processed) 'IMG_20150730_115645_3' -> 2.6636 secs	-'(memoized) 'IMG_20150730_115645_3' -> 0.1500 secs
-'(processed) 'IMG_20150730_115645_4' -> 2.6275 secs	-'(memoized) 'IMG_20150730_115645_4' -> 0.1451 secs
-'(processed) 'IMG_20150730_115646' -> 2.5770 secs	-'(memoized) 'IMG_20150730_115646' -> 0.1902 secs
-'(processed) 'IMG_20150730_115646_1' -> 2.6437 secs	-'(memoized) 'IMG_20150730_115646_1' -> 0.2722 secs
-'(processed) 'IMG_20150730_115646_2' -> 2.6496 secs	-'(memoized) 'IMG_20150730_115646_2' -> 0.1825 secs
-'(processed) 'IMG_20150730_115646_3' -> 2.7298 secs	-'(memoized) 'IMG_20150730_115646_3' -> 0.2998 secs
-'(processed) 'IMG_20150730_115646_5' -> 2.7782 secs	-'(memoized) 'IMG_20150730_115646_5' -> 0.2395 secs
-'(processed) 'IMG_20150730_115647' -> 2.8211 secs	-'(memoized) 'IMG_20150730_115647' -> 0.1999 secs
-'(processed) 'IMG_20150730_115647_2' -> 2.7638 secs	-'(memoized) 'IMG_20150730_115647_2' -> 0.2818 secs
-'(processed) 'IMG_20150730_115647_3' -> 2.6771 secs	-'(memoized) 'IMG_20150730_115647_3' -> 0.2831 secs
-'(processed) 'IMG_20150730_115647_4' -> 2.7531 secs	-'(memoized) 'IMG_20150730_115647_4' -> 0.5242 secs
-'(processed) 'IMG_20150730_115648' -> 2.7393 secs	-'(memoized) 'IMG_20150730_115648' -> 0.1178 secs
-'(processed) 'IMG_20150730_115648_1' -> 2.7187 secs	-'(memoized) 'IMG_20150730_115648_1' -> 0.1983 secs
-'(processed) 'IMG_20150730_115648_2' -> 2.7531 secs	-'(memoized) 'IMG_20150730_115648_2' -> 0.2691 secs
-'(processed) 'IMG_20150730_115648_3' -> 2.8693 secs	-'(memoized) 'IMG_20150730_115648_3' -> 0.1850 secs
-'(processed) 'IMG_20150730_115648_4' -> 2.6890 secs	-'(memoized) 'IMG_20150730_115648_4' -> 0.2002 secs
-'(processed) 'IMG_20150730_115649' -> 2.9883 secs	-'(memoized) 'IMG_20150730_115649' -> 0.1064 secs
-'(processed) 'IMG_20150730_115649_1' -> 3.1665 secs	-'(memoized) 'IMG_20150730_115649_1' -> 0.1482 secs
-'(processed) 'IMG_20150730_115649_2' -> 2.7142 secs	-'(memoized) 'IMG_20150730_115649_2' -> 0.3206 secs
-'(processed) 'IMG_20150730_115649_4' -> 2.7838 secs	-'(memoized) 'IMG_20150730_115649_4' -> 0.2173 secs
-'(processed) 'IMG_20150730_115650' -> 2.8794 secs	-'(memoized) 'IMG_20150730_115650' -> 0.1346 secs
-'(processed) 'IMG_20150730_115650_3' -> 2.8313 secs	-'(memoized) 'IMG_20150730_115650_3' -> 0.2424 secs
-'(processed) 'IMG_20150730_115650_4' -> 2.7142 secs	-'(memoized) 'IMG_20150730_115650_4' -> 0.2487 secs
-'(processed) 'IMG_20150730_115651' -> 2.8182 secs	-'(memoized) 'IMG_20150730_115651' -> 0.3232 secs
-'Restoring' -> 33.9889 secs	-'Restoring' -> 83.8567 secs
-'Matching' -> 1.8534 secs	-'Matching' -> 4.6324 secs
-'Merging' -> 32.1306 secs	-'Merging' -> 79.2189 secs
-''IMG_20150730_115646_2' -> 1.4884 secs	-''IMG_20150730_115646_2' -> 2.5882 secs
-''IMG_20150730_115646_1' -> 1.3321 secs	-''IMG_20150730_115646_1' -> 3.2101 secs
-''IMG_20150730_115646' -> 1.2919 secs	-''IMG_20150730_115646' -> 2.5621 secs
-''IMG_20150730_115645_4' -> 1.1619 secs	-''IMG_20150730_115645_4' -> 3.2147 secs
-''IMG_20150730_115645_3' -> 1.2057 secs	-''IMG_20150730_115645_3' -> 3.4402 secs
-''IMG_20150730_115646_5' -> 1.2746 secs	-''IMG_20150730_115646_5' -> 4.1303 secs
-''IMG_20150730_115647_1' -> 1.2460 secs	-''IMG_20150730_115647_1' -> 3.2987 secs
-''IMG_20150730_115647' -> 1.2818 secs	-''IMG_20150730_115647' -> 3.0160 secs
-''IMG_20150730_115647_2' -> 1.2583 secs	-''IMG_20150730_115647_2' -> 2.3737 secs
-''IMG_20150730_115647_3' -> 1.2666 secs	-''IMG_20150730_115647_3' -> 2.4315 secs
-''IMG_20150730_115647_4' -> 1.2700 secs	-''IMG_20150730_115647_4' -> 3.5084 secs
-''IMG_20150730_115648_1' -> 1.3461 secs	-''IMG_20150730_115648_1' -> 3.6602 secs
-''IMG_20150730_115648' -> 1.2728 secs	-''IMG_20150730_115648' -> 3.4772 secs
-''IMG_20150730_115651' -> 1.4752 secs	-''IMG_20150730_115651' -> 4.4735 secs
-''IMG_20150730_115650_4' -> 1.4091 secs	-''IMG_20150730_115650_4' -> 3.0858 secs
-''IMG_20150730_115650_3' -> 1.3400 secs	-''IMG_20150730_115650_3' -> 4.2219 secs
-''IMG_20150730_115648_2' -> 1.3298 secs	-''IMG_20150730_115648_2' -> 3.6904 secs
-''IMG_20150730_115648_3' -> 1.4255 secs	-''IMG_20150730_115648_3' -> 3.4017 secs
-''IMG_20150730_115648_4' -> 1.4204 secs	-''IMG_20150730_115648_4' -> 3.3823 secs
-''IMG_20150730_115649' -> 1.3792 secs	-''IMG_20150730_115649' -> 3.4723 secs
-''IMG_20150730_115649_2' -> 1.4441 secs	-''IMG_20150730_115649_2' -> 3.3438 secs
-''IMG_20150730_115649_1' -> 1.4099 secs	-''IMG_20150730_115649_1' -> 3.0954 secs
-''IMG_20150730_115649_4' -> 1.3202 secs	-''IMG_20150730_115649_4' -> 3.3002 secs
-''IMG_20150730_115650' -> 1.3939 secs	-''IMG_20150730_115650' -> 2.6352 secs
-'Post-processing' -> 0.4401 secs	-'Post-processing' -> 0.9259 secs

Figure 72. Result for set19 using command ‘imrestore ./results/set19/*.* --lens --overwrite --cachePath {temp}’



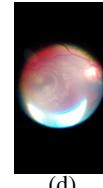
(a)



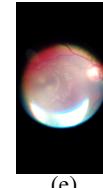
(b)



(c)

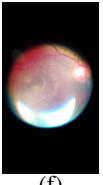


(d)

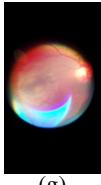


(e)

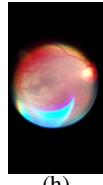
IMG_20150730_115646_3IMG_20150730_115646_2IMG_20150730_115646_1IMG_20150730_115646 IMG_20150730_115645_4



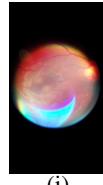
(f)



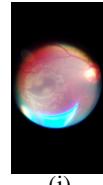
(g)



(h)

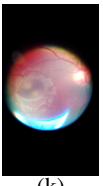


(i)

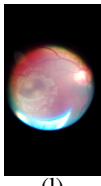


(j)

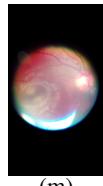
IMG_20150730_115645_3IMG_20150730_115646_5IMG_20150730_115647_1IMG_20150730_115647 IMG_20150730_115647_2



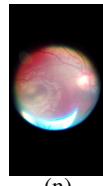
(k)



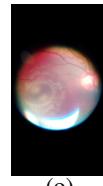
(l)



(m)

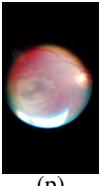


(n)

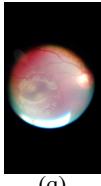


(o)

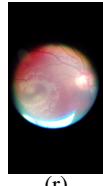
IMG_20150730_115647_3IMG_20150730_115647_4IMG_20150730_115648_1IMG_20150730_115648 IMG_20150730_115651



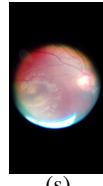
(p)



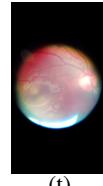
(q)



(r)

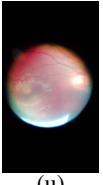


(s)

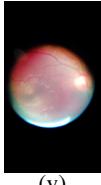


(t)

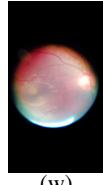
IMG_20150730_115650_4IMG_20150730_115650_3IMG_20150730_115648_2IMG_20150730_115648_3IMG_20150730_115648_4



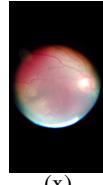
(u)



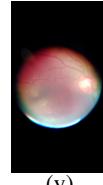
(v)



(w)



(x)



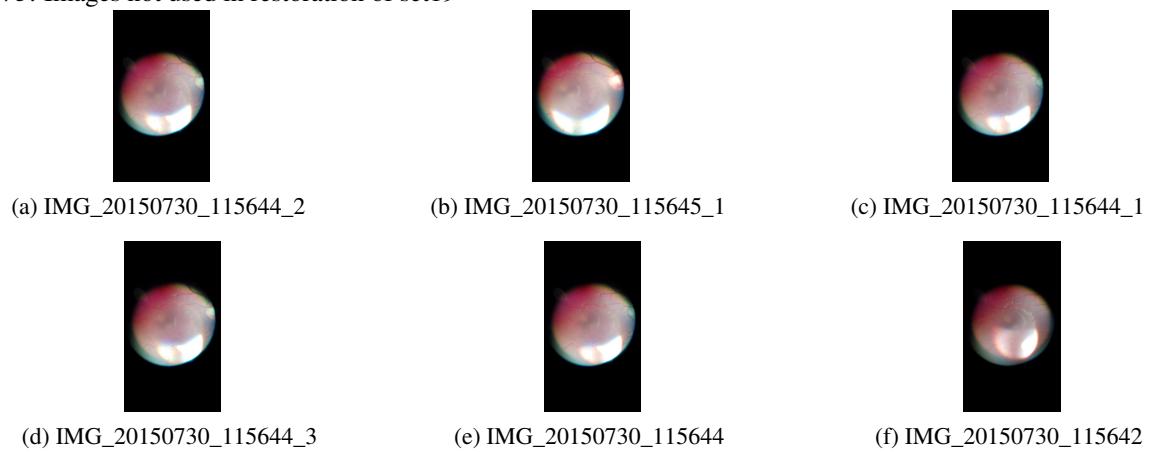
(y)

IMG_20150730_115649 IMG_20150730_115649_2IMG_20150730_115649_1IMG_20150730_115649_4IMG_20150730_115650



(z) Restored image

Figure 73. Images not used in restoration of set19



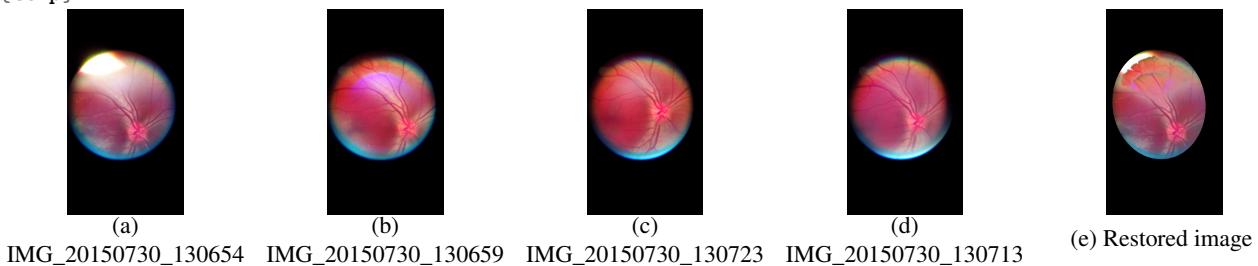
Result for set 20

Table 26

Profiling for set20 using command ‘imrestore ../results/set20/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 54.0407 secs	'ImRestore init' -> 26.4929 secs
└─ 'Computing features' -> 50.1011 secs	└─ 'Computing features' -> 9.7973 secs
└─ (processed) '20150730_132852' -> 2.4145 secs	└─ (memoized) '20150730_132852' -> 0.2608 secs
└─ (processed) '20150730_132857' -> 2.0729 secs	└─ (memoized) '20150730_132857' -> 0.1316 secs
└─ (processed) '20150730_132900' -> 2.3489 secs	└─ (memoized) '20150730_132900' -> 0.3009 secs
└─ (processed) '20150730_132904' -> 2.0129 secs	└─ (memoized) '20150730_132904' -> 0.2884 secs
└─ (processed) '20150730_132917' -> 1.7671 secs	└─ (memoized) '20150730_132917' -> 0.0986 secs
└─ (processed) '20150730_132924' -> 1.9242 secs	└─ (memoized) '20150730_132924' -> 0.3143 secs
└─ (processed) '20150730_132929' -> 1.8181 secs	└─ (memoized) '20150730_132929' -> 0.5799 secs
└─ (processed) '20150730_132932' -> 1.8655 secs	└─ (memoized) '20150730_132932' -> 0.3983 secs
└─ (processed) '20150730_132936' -> 1.7875 secs	└─ (memoized) '20150730_132936' -> 0.3612 secs
└─ (processed) 'IMG_20150730_122502' -> 1.7832 secs	└─ (memoized) 'IMG_20150730_122502' -> 0.4372 secs
└─ (processed) 'IMG_20150730_122504' -> 1.8384 secs	└─ (memoized) 'IMG_20150730_122504' -> 0.2810 secs
└─ (processed) 'IMG_20150730_122506' -> 1.8803 secs	└─ (memoized) 'IMG_20150730_122506' -> 0.2850 secs
└─ (processed) 'IMG_20150730_122520' -> 1.7905 secs	└─ (memoized) 'IMG_20150730_122520' -> 0.3053 secs
└─ (processed) 'IMG_20150730_130654' -> 2.0385 secs	└─ (memoized) 'IMG_20150730_130654' -> 0.4560 secs
└─ (processed) 'IMG_20150730_130659' -> 2.1586 secs	└─ (memoized) 'IMG_20150730_130659' -> 0.4691 secs
└─ (processed) 'IMG_20150730_130706' -> 2.3120 secs	└─ (memoized) 'IMG_20150730_130706' -> 0.3788 secs
└─ (processed) 'IMG_20150730_130713' -> 2.2099 secs	└─ (memoized) 'IMG_20150730_130713' -> 0.5561 secs
└─ (processed) 'IMG_20150730_130719' -> 2.0846 secs	└─ (memoized) 'IMG_20150730_130719' -> 0.4428 secs
└─ (processed) 'IMG_20150730_130723' -> 2.4502 secs	└─ (memoized) 'IMG_20150730_130723' -> 1.1865 secs
└─ (processed) 'IMG_20150730_130729' -> 2.3709 secs	└─ (memoized) 'IMG_20150730_130729' -> 0.7323 secs
└─ (processed) 'IMG_20150730_130756' -> 2.2615 secs	└─ (memoized) 'IMG_20150730_130756' -> 0.2676 secs
└─ (processed) 'IMG_20150730_130759' -> 2.0909 secs	└─ (memoized) 'IMG_20150730_130759' -> 0.1314 secs
└─ (processed) 'IMG_20150730_130803' -> 2.0534 secs	└─ (memoized) 'IMG_20150730_130803' -> 0.2384 secs
└─ (processed) 'IMG_20150730_130809' -> 1.8940 secs	└─ (memoized) 'IMG_20150730_130809' -> 0.2800 secs
└─ 'Restoring' -> 3.6517 secs	└─ 'Restoring' -> 14.8907 secs
└─ 'Matching' -> 0.4430 secs	└─ 'Matching' -> 1.6175 secs
└─ 'Merging' -> 3.2080 secs	└─ 'Merging' -> 13.2723 secs
└─ 'IMG_20150730_130659' -> 0.9521 secs	└─ 'IMG_20150730_130659' -> 4.0912 secs
└─ 'IMG_20150730_130723' -> 0.9611 secs	└─ 'IMG_20150730_130723' -> 4.0746 secs
└─ 'IMG_20150730_130713' -> 1.1127 secs	└─ 'IMG_20150730_130713' -> 4.3786 secs
└─ 'Post-processing' -> 0.2879 secs	└─ 'Post-processing' -> 1.8048 secs

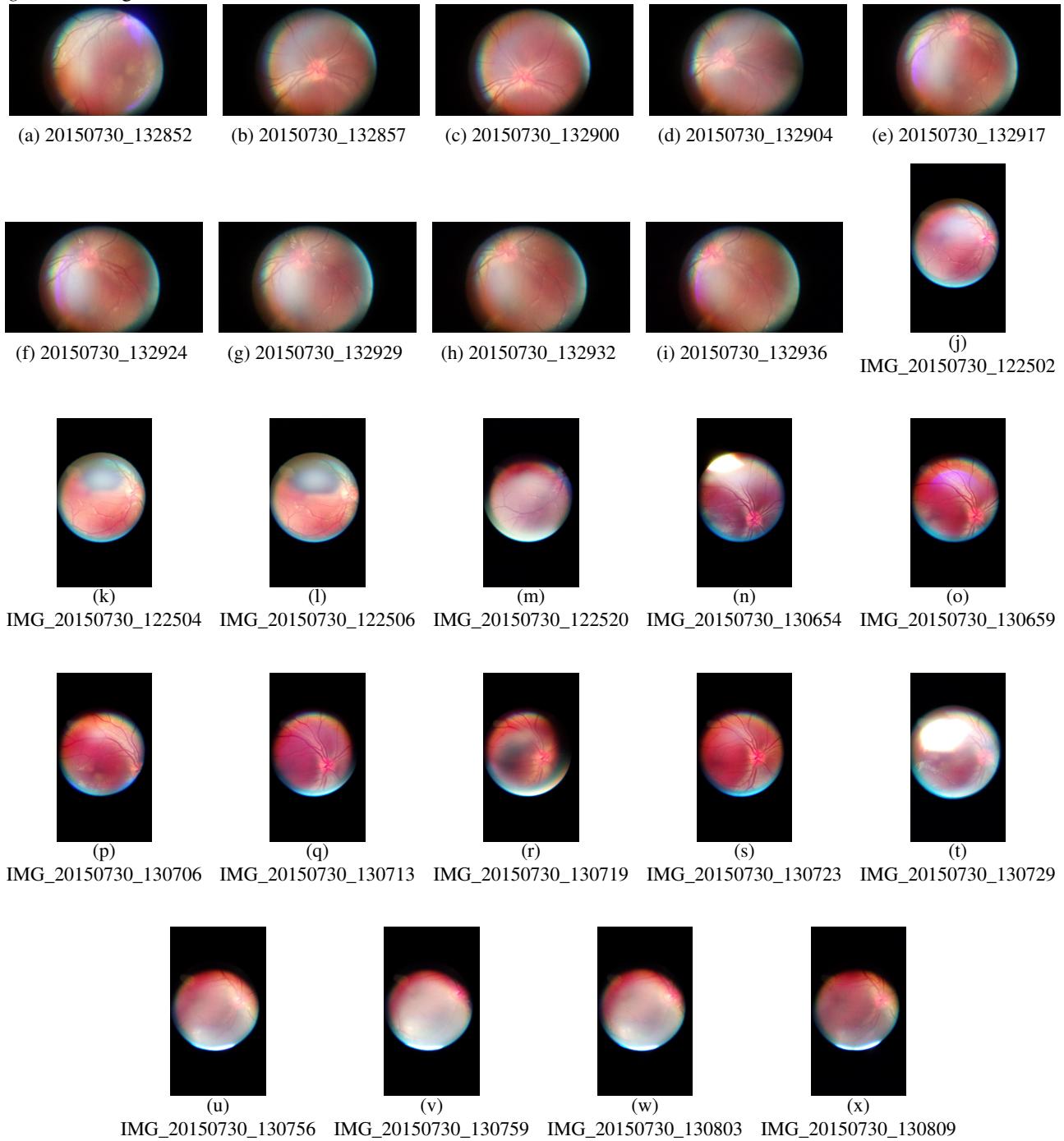
Figure 74. Result for set20 using command ‘imrestore ../results/set20/*.* --lens --overwrite --cachePath {temp}’



Using Code 5 in set20 *imrestore* processed 4 of the 24 images in set20, this is shown in Figure 75. 3 is the number of images merged in the *base image*. 20 is the number of images not used for the set20. Figure 74 presents the results for set20. ?? shows the images not used in

set20. Figure 74a shows the *base image*, Figure 74e shows the restored image and the remaining merged images. Figure 75a shows the first image in set. ?? shows the first image not used in the merging. Table 26 presents the profiling produced for set20. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 54.04 to 26.49 seconds saving in total 27.55 seconds. This is translated in a gain of the 50.98% with cache time being 49.02% and this means that an excess in time of the 203.98% would happen if passed from cached to a not cached session the rate of used images is 0.17. the rate of merged images is 0.12. the rate of failed images is 0.83.

Figure 75. Images in set20



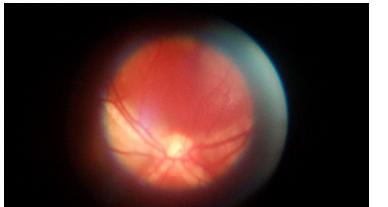
Result for set 21

Table 27

Profiling for set21 using command ‘imrestore ../results/set21/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 29.1726 secs	'ImRestore init' -> 2.8818 secs
└─ 'Computing features' -> 28.4758 secs	└─ 'Computing features' -> 1.7148 secs
└─ (processed) '20150730_131816' -> 2.3519 secs	└─ (memoized) '20150730_131816' -> 0.4360 secs
└─ (processed) '20150730_131823' -> 2.3904 secs	└─ (memoized) '20150730_131823' -> 0.2340 secs
└─ (processed) '20150730_131827' -> 2.0813 secs	└─ (memoized) '20150730_131827' -> 0.3120 secs
└─ (processed) '20150730_131834' -> 2.9239 secs	└─ (memoized) '20150730_131834' -> 0.1405 secs
└─ (processed) '20150730_131836' -> 2.7180 secs	└─ (memoized) '20150730_131836' -> 0.1112 secs
└─ (processed) '20150730_131839' -> 2.9072 secs	└─ (memoized) '20150730_131839' -> 0.1017 secs
└─ (processed) '20150730_131844' -> 3.6443 secs	└─ (memoized) '20150730_131844' -> 0.0799 secs
└─ (processed) '20150730_131913' -> 3.1464 secs	└─ (memoized) '20150730_131913' -> 0.0542 secs
└─ (processed) '20150730_131915' -> 2.8072 secs	└─ (memoized) '20150730_131915' -> 0.0737 secs
└─ (processed) '20150730_131916' -> 2.9230 secs	└─ (memoized) '20150730_131916' -> 0.0703 secs
└─ 'Restoring' -> 0.0880 secs	└─ 'Restoring' -> 0.2236 secs
└─ 'Matching' -> 0.0340 secs	└─ 'Matching' -> 0.0542 secs
└─ 'Merging' -> 0.0535 secs	└─ 'Merging' -> 0.1691 secs
└─ 'Post-processing' -> 0.6087 secs	└─ 'Post-processing' -> 0.9434 secs

Figure 76. Result for set21 using command ‘imrestore ../results/set21/*.* --lens --overwrite --cachePath {temp}’

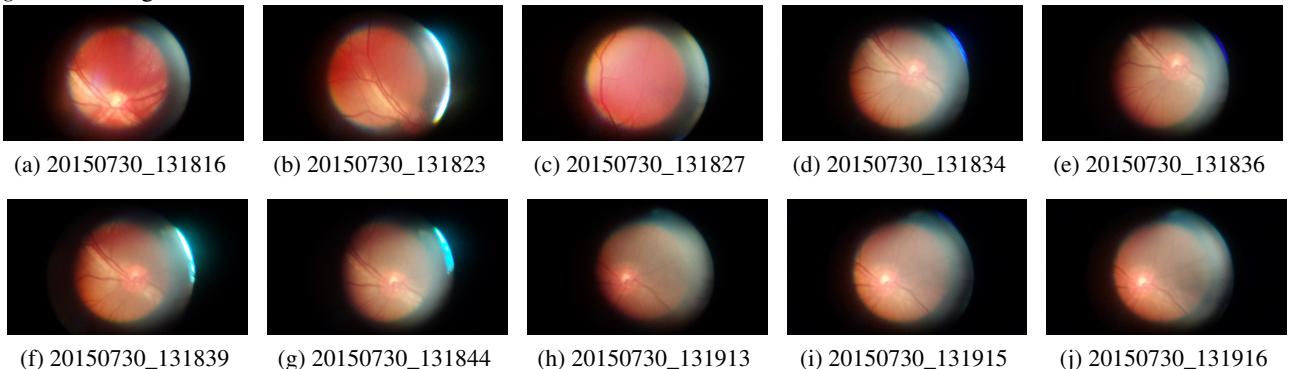


(a) 20150730_131816



(b) Restored image

Figure 77. Images in set21



Using Code 5 in set21 *imrestore* processed 1 of the 10 images in set21, this is shown in Figure 77. 0 is the number of images merged in the *base image*. 9 is the number of images not used for the set21. Figure 76 presents the results for set21. ?? shows the images not used in set21. Figure 76a shows the *base image*, Figure 76b shows the restored image and the remaining merged image. Figure 77a shows the first image in set. ?? shows the first image not used in the merging. Table 27 presents the profiling produced for set21. It is evidenced that the

time was reduced by 10 times of the uncached session. In this way time passed from 29.17 to 2.88 seconds saving in total 26.29 seconds. This is translated in a gain of the 90.12% with cache time being 9.88% and this means that an excess in time of the 1012.29% would happen if passed from cached to a not cached session the rate of used images is 0.10. the rate of merged images is 0.00. the rate of failed images is 0.90.

Result for set 22

Table 28

Profiling for set22 using command ‘imrestore ../results/set22/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 21.4743 secs	'ImRestore init' -> 12.1105 secs
└─ 'Computing features' -> 17.9448 secs	└─ 'Computing features' -> 1.9042 secs
└─ (processed) 'IMG_20150730_121750' -> 2.6784 secs	└─ (memoized) 'IMG_20150730_121750' -> 0.2615 secs
└─ (processed) 'IMG_20150730_121756' -> 2.8147 secs	└─ (memoized) 'IMG_20150730_121756' -> 0.3144 secs
└─ (processed) 'IMG_20150730_121801' -> 2.9608 secs	└─ (memoized) 'IMG_20150730_121801' -> 0.3144 secs
└─ (processed) 'IMG_20150730_121804' -> 2.9499 secs	└─ (memoized) 'IMG_20150730_121804' -> 0.2827 secs
└─ (processed) 'IMG_20150730_121804_1' -> 3.2179 secs	└─ (memoized) 'IMG_20150730_121804_1' -> 0.3336 secs
└─ (processed) 'IMG_20150730_121804_2' -> 3.0732 secs	└─ (memoized) 'IMG_20150730_121804_2' -> 0.2384 secs
└─ 'Restoring' -> 3.1301 secs	└─ 'Restoring' -> 8.9032 secs
└─ 'Matching' -> 0.2217 secs	└─ 'Matching' -> 0.4129 secs
└─ 'Merging' -> 2.9078 secs	└─ 'Merging' -> 8.4896 secs
└─ 'IMG_20150730_121801' -> 4.1115 secs	└─ 'IMG_20150730_121801' -> 4.1347 secs
└─ 'IMG_20150730_121750' -> 1.4492 secs	└─ 'IMG_20150730_121750' -> 4.1933 secs
└─ 'Post-processing' -> 0.3994 secs	└─ 'Post-processing' -> 1.3031 secs

Figure 78. Result for set22 using command ‘imrestore ../results/set22/*.* --lens --overwrite --cachePath {temp}’,



(a) IMG_20150730_121756 (b) IMG_20150730_121801 (c) IMG_20150730_121750 (d) Restored image

Figure 79. Images not used in restoration of set22



(a) IMG_20150730_121804_2 (b) IMG_20150730_121804 (c) IMG_20150730_121804_1

Using Code 5 in set22 *imrestore* processed 3 of the 6 images in set22, this is shown in ???. 2 is the number of images merged in the *base image*. 3 is the number of images not used for the set22. Figure 78 presents the results for set22. Figure 79 shows the images not used in set22. Figure 78a shows the *base image*, Figure 78d shows the restored image and the remaining merged images. ?? shows the first image in set. Figure 79a shows the first image not used in the merging. Table 28 presents the profiling produced for set22. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 21.47 to 12.11 seconds saving in total 9.36 seconds. This is translated in a gain of the 43.60% with cache time being 56.40% and this means that an excess in time of the 177.32% would happen if passed from cached to a not cached session the rate of used images is 0.50. the rate of merged images is 0.33. the rate of failed images is 0.50.

Result for set 23

Table 29

Profiling for set23 using command ‘imrestore ../results/set23/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 27.8534 secs	'ImRestore init' -> 4.0939 secs
'Computing features' -> 27.2165 secs	'Computing features' -> 2.7341 secs
(processed) 'IMG_20150730_122140' -> 3.2541 secs	(memoized) 'IMG_20150730_122140' -> 0.2125 secs
(processed) 'IMG_20150730_122154' -> 3.0911 secs	(memoized) 'IMG_20150730_122154' -> 0.2942 secs
(processed) 'IMG_20150730_122200' -> 3.0967 secs	(memoized) 'IMG_20150730_122200' -> 0.5929 secs
(processed) 'IMG_20150730_122216' -> 2.9525 secs	(memoized) 'IMG_20150730_122216' -> 0.3233 secs
(processed) 'IMG_20150730_122225' -> 2.8391 secs	(memoized) 'IMG_20150730_122225' -> 0.3009 secs
(processed) 'IMG_20150730_122227' -> 2.8896 secs	(memoized) 'IMG_20150730_122227' -> 0.3483 secs
(processed) 'IMG_20150730_122229' -> 2.8071 secs	(memoized) 'IMG_20150730_122229' -> 0.2184 secs
(processed) 'IMG_20150730_122235' -> 2.8947 secs	(memoized) 'IMG_20150730_122235' -> 0.1439 secs
(processed) 'IMG_20150730_122237' -> 2.9100 secs	(memoized) 'IMG_20150730_122237' -> 0.1124 secs
'Restoring' -> 0.1615 secs	'Restoring' -> 0.4669 secs
'Matching' -> 0.0795 secs	'Matching' -> 0.1127 secs
'Merging' -> 0.0818 secs	'Merging' -> 0.3540 secs
'Post-processing' -> 0.4754 secs	'Post-processing' -> 0.8929 secs

Figure 80. Result for set23 using command ‘imrestore ../results/set23/*.* --lens --overwrite --cachePath {temp}’

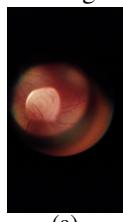


(a) IMG_20150730_122200

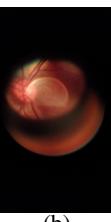
(b) Restored image

Using Code 5 in set23 *imrestore* processed 1 of the 9 images in set23, this is shown in Figure 81. 0 is the number of images merged in the *base image*. 8 is the number of images not used for the set23. Figure 80 presents the results for set23. ?? shows the images not used in set23. Figure 80a shows the *base image*, Figure 80b shows the restored image and the remaining merged image. Figure 81a shows the first image in set. ?? shows the first image not used in the merging. Table 29 presents the profiling produced for set23. It is evidenced that the time was reduced by 7 times of the uncached session. In this way time passed from 27.85 to 4.09 seconds saving in total 23.76 seconds. This is translated in a gain of the 85.30% with cache time being 14.70% and this means that an excess in time of the 680.36% would happen if passed from cached to a not cached session the rate of used images is 0.11. the rate of merged images is 0.00. the rate of failed images is 0.89.

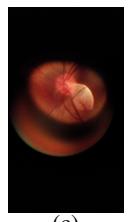
Figure 81. Images in set23



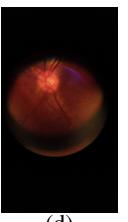
(a)



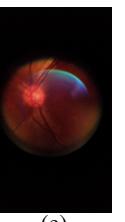
(b)



(c)

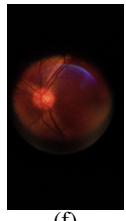


(d)

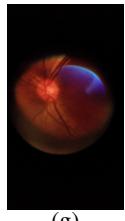


(e)

IMG_20150730_122140 IMG_20150730_122154 IMG_20150730_122200 IMG_20150730_122216 IMG_20150730_122225



(f)



(g)



(h)



(i)

IMG_20150730_122227 IMG_20150730_122229 IMG_20150730_122235 IMG_20150730_122237

Result for set 24

Table 30

Profiling for set24 using command ‘imrestore/results/set24/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 17.4084 secs	'ImRestore init' -> 6.2216 secs
└─ 'Computing features' -> 13.7685 secs	└─ 'Computing features' -> 1.4231 secs
└─ (processed) 'IMG_20150730_125212' -> 3.6106 secs	└─ (memoized) 'IMG_20150730_125212' -> 0.3641 secs
└─ (processed) 'IMG_20150730_125216' -> 3.4603 secs	└─ (memoized) 'IMG_20150730_125216' -> 0.4247 secs
└─ (processed) 'IMG_20150730_125222' -> 3.3301 secs	└─ (memoized) 'IMG_20150730_125222' -> 0.3686 secs
└─ (processed) 'IMG_20150730_125228' -> 3.1819 secs	└─ (memoized) 'IMG_20150730_125228' -> 0.1722 secs
└─ 'Restoring' -> 3.2164 secs	└─ 'Restoring' -> 3.7034 secs
└─ 'Matching' -> 0.0737 secs	└─ 'Matching' -> 0.1339 secs
└─ 'Merging' -> 3.1420 secs	└─ 'Merging' -> 3.5691 secs
└─ 'IMG_20150730_125212' -> 1.5209 secs	└─ 'IMG_20150730_125212' -> 3.4510 secs
└─ 'IMG_20150730_125222' -> 1.5719 secs	
└─ 'Post-processing' -> 0.4235 secs	└─ 'Post-processing' -> 1.0951 secs

Figure 82. Result for set24 using command ‘imrestore/results/set24/*.* --lens --overwrite --cachePath {temp}’



(a) IMG_20150730_125216

(b) IMG_20150730_125212

(c) Restored image

Figure 83. Images in set24



(a) IMG_20150730_125212

(b) IMG_20150730_125216

(c) IMG_20150730_125222

(d) IMG_20150730_125228

Using Code 5 in set24 *imrestore* processed 2 of the 4 images in set24, this is shown in Figure 83. 1 is the number of images merged in the *base image*. 2 is the number of images not used for the set24. Figure 82 presents the results for set24. ?? shows the images not used in set24. Figure 82a shows the *base image*, Figure 82c shows the restored image and the remaining merged images. Figure 83a shows the first image in set. ?? shows the first image not used in the merging. Table 30 presents the profiling produced for set24. It is evidenced that the time was reduced by 3 times of the uncached session. In this way time passed from 17.41 to 6.22 seconds saving in total 11.19 seconds. This is translated in a gain of the 64.26% with cache time being 35.74% and this means that an excess in time of the 279.81% would happen if passed from cached to a not cached session the rate of used images is 0.50. the rate of merged images is 0.25. the rate of failed images is 0.50.

Result for set 25

Table 31

Profiling for set25 using command ‘imrestore .../results/set25/.* --lens --overwrite --cachePath {temp}’*

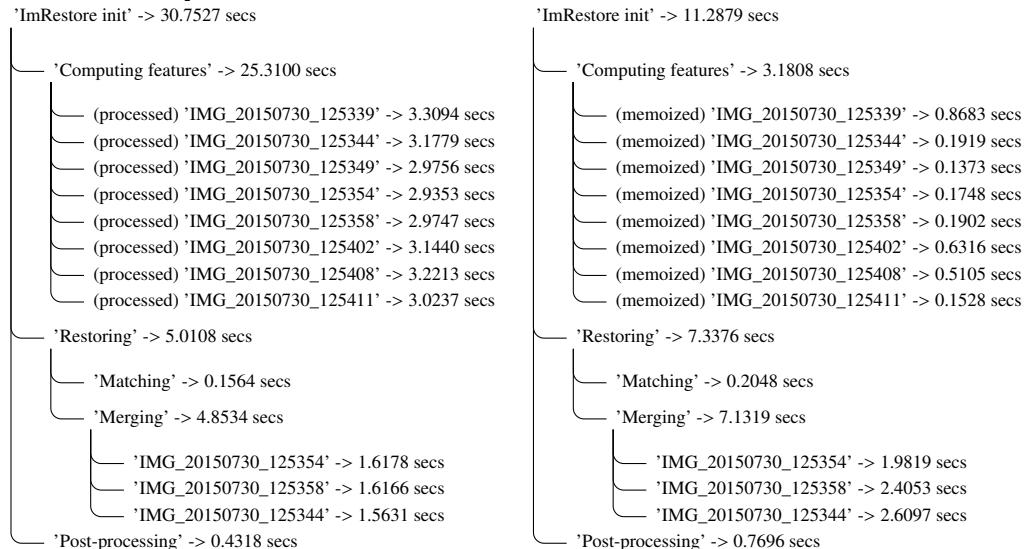


Figure 84. Result for set25 using command ‘imrestore .../results/set25/*.* --lens --overwrite --cachePath {temp}’

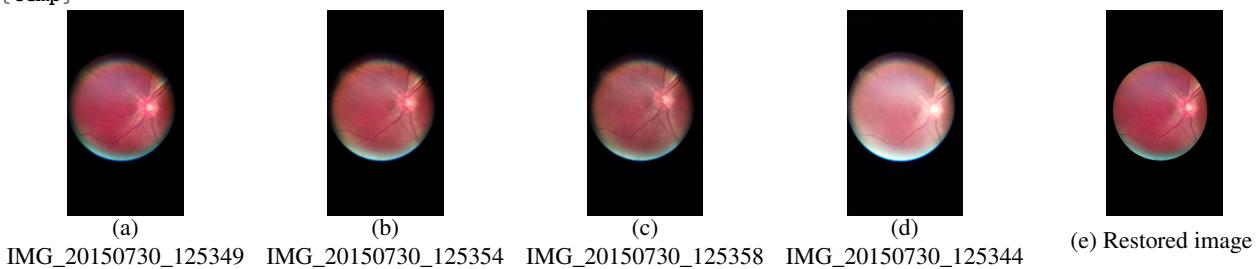
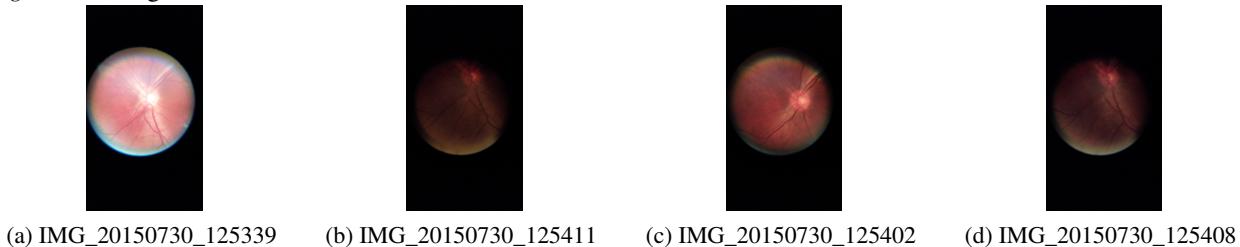


Figure 85. Images not used in restoration of set25



Using Code 5 in set25 *imrestore* processed 4 of the 8 images in set25, this is shown in ???. 3 is the number of images merged in the *base image*. 4 is the number of images not used for the set25. Figure 84 presents the results for set25. Figure 85 shows the images not used in set25. Figure 84a shows the *base image*, Figure 84e shows the restored image and the remaining merged images. ?? shows the first image in set. Figure 85a shows the first image not used in the merging. Table 31 presents the profiling produced for set25. It is evidenced that the time was reduced by 3 times of the uncached session. In this way time passed from 30.75 to 11.29 seconds saving in total 19.46 seconds.

This is translated in a gain of the 63.29% with cache time being 36.71% and this means that an excess in time of the 272.44% would happen if passed from cached to a not cached session the rate of used images is 0.50. the rate of merged images is 0.38. the rate of failed images is 0.50.

Result for set 26

Table 32

Profiling for set26 using command ‘imrestore ../results/set26/.* --lens --overwrite --cachePath {temp}’*

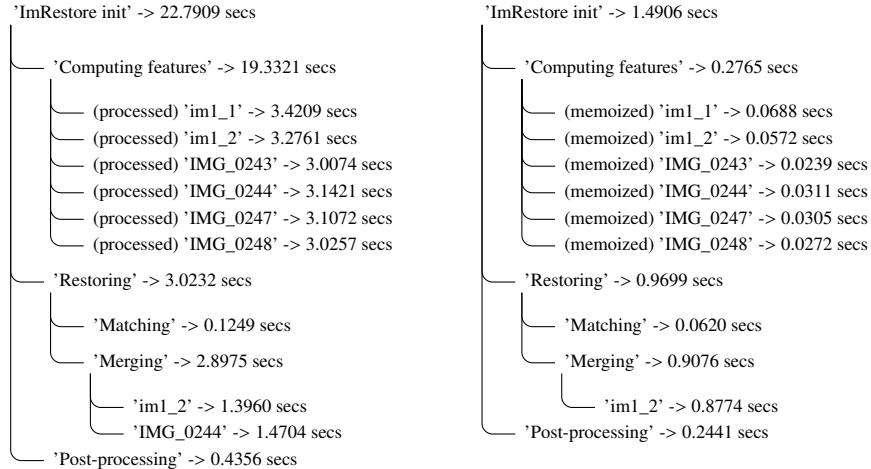


Figure 86. Result for set26 using command ‘imrestore ../results/set26/*.* --lens --overwrite --cachePath {temp}’

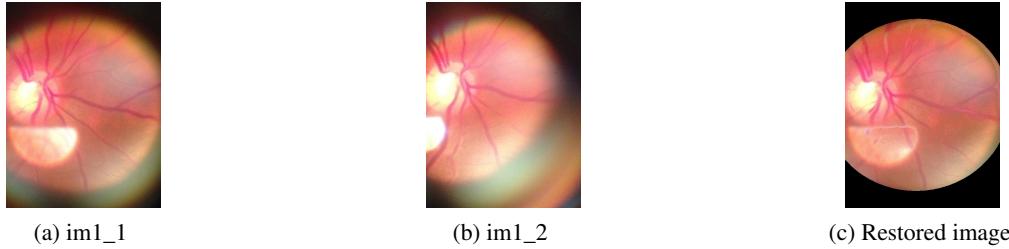
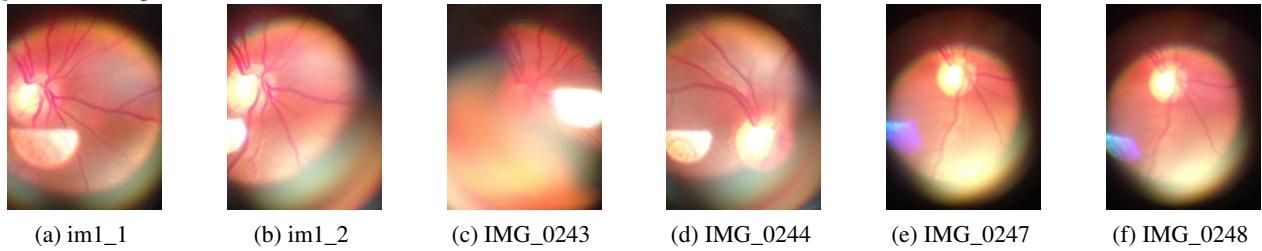


Figure 87. Images in set26



Using Code 5 in set26 *imrestore* processed 2 of the 6 images in set26, this is shown in Figure 87. 1 is the number of images merged in the *base image*. 4 is the number of images not used for the set26. Figure 86 presents the results for set26. ?? shows the images not used in set26. Figure 86a shows the *base image*, Figure 86c shows the restored image and the remaining merged images. Figure 87a shows the first image in set. ?? shows the first image not used in the merging. Table 32 presents the profiling produced for set26. It is evidenced that the time was reduced by 15 times of the uncached session. In this way time passed from 22.79 to 1.49 seconds saving in total 21.30 seconds. This is translated in a gain of the 93.46% with cache time being 6.54% and this means that an excess in time of the 1529.02% would happen if passed from cached to a not cached session the rate of used images is 0.33. the rate of merged images is 0.17. the rate of failed images is 0.67.

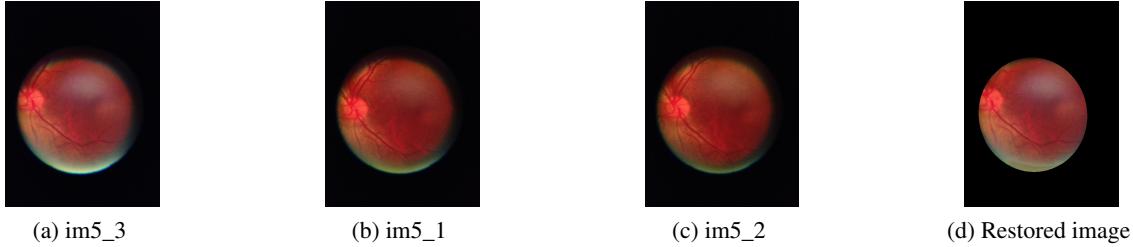
Result for set 27

Table 33

Profiling for set27 using command ‘imrestore ../results/set27/.* --lens --overwrite --cachePath {temp}’*

'ImRestore init' -> 13.4935 secs	'ImRestore init' -> 8.8181 secs
└─ 'Computing features' -> 9.7324 secs	└─ 'Computing features' -> 0.8755 secs
└─ (processed) 'im5_1' -> 3.4046 secs	└─ (memoized) 'im5_1' -> 0.3627 secs
└─ (processed) 'im5_2' -> 3.1560 secs	└─ (memoized) 'im5_2' -> 0.1645 secs
└─ (processed) 'im5_3' -> 3.0355 secs	└─ (memoized) 'im5_3' -> 0.2879 secs
└─ 'Restoring' -> 3.2621 secs	└─ 'Restoring' -> 7.0875 secs
└─ 'Matching' -> 0.0385 secs	└─ 'Matching' -> 0.0842 secs
└─ 'Merging' -> 3.2232 secs	└─ 'Merging' -> 6.9966 secs
└─ 'im5_1' -> 1.6538 secs	└─ 'im5_1' -> 3.6402 secs
└─ 'im5_2' -> 1.5693 secs	└─ 'im5_2' -> 3.3563 secs
└─ 'Post-processing' -> 0.4991 secs	└─ 'Post-processing' -> 0.8551 secs

Figure 88. Result for set27 using command ‘imrestore ../results/set27/*.* --lens --overwrite --cachePath {temp}’



Using Code 5 in set27 *imrestore* processed 3 of the 3 images in set27, this is shown in ???. 2 is the number of images merged in the *base image*. 0 is the number of images not used for the set27. Figure 88 presents the results for set27. ?? shows the images not used in set27. Figure 88a shows the *base image*, Figure 88d shows the restored image and the remaining merged images. ?? shows the first image not used in the merging. Table 33 presents the profiling produced for set27. It is evidenced that the time was reduced by 2 times of the uncached session. In this way time passed from 13.49 to 8.82 seconds saving in total 4.68 seconds. This is translated in a gain of the 34.65% with cache time being 65.35% and this means that an excess in time of the 153.02% would happen if passed from cached to a not cached session the rate of used images is 1.00. the rate of merged images is 0.67. the rate of failed images is 0.00.

Validation

Discussion

Illumination is not a big issue whenever using descriptor methods like SIFT, SURF or ORB (see Figure 84) to match images but when these are merged they tend to have problems of color difference whenever a color is more vivid than the other producing unappealing results (see Figure 65 and Figure 70). A direct solution is to apply histogram matching so that one of the merging image’s color approaches to the other, this solves the problem well when they have similar lighting conditions but begin to be noticeable when not. Then it is recommended that the user takes the set of images at similar lighting conditions.

There are factors that influence in the computation of key-points and matching like for example directly filtering desired key-points by applying masks for desired areas or adjusting the feature detector parameters. Though image resizing produce matching between images it is not guarantee that they will be merged equally when they are of different sizes. Also, blurriness in general affects the identification of candidate key-points. This is demonstrated in the application of filters that tend to produce blurry results. Opposite to this, color adjustment methods that do not produce consistent smooth results can create false points of interest that do not correspond to the original image scenery.

Because the alpha mask is calculated at each merging it depends on the previous results causing errors to be cumulative. Some of the solutions can be to keep track of the changes made to the merging image so that these can be taken into consideration when calculating the alpha mask. Of course this can lead to performance and memory penalties, additionally to making the code more complicated. These errors are being reduced with the improvement of the segmentation algorithm identifying regions to correct and leaving correct areas untouched. More is compensated using alpha masks instead of binary mask to produce smooth operations in segmentation procedures.

The restoration algorithm lacks the capability of classifying good and bad images according to a case, it just ranks them according to a evaluation criteria. Said cases like in section Result for set 6 hinders the capabilities of the restoration due that the algorithm does not checks for unwanted images and it is responsibility of the user. In fact, for retinal restoration the user could feed in images that not correspond to retinal areas and still they would be processed.

Conclusions

Many algorithms were investigated to see if they could be of any use for the project (see Algorithms design section). Those that were promising where tested and further developed for the realization of partial goals in the project dividing the problem into several stages that could be processed to get near the desired application.

The developed algorithms were validated with their corresponding theory and in some cases comparisons were carried out between their estimations and the desired outputs so that they could offer a gain in the processing of images.

These tools were organized according to their category and packed into the *RRtoolbox* package and then used to create the *imrestore* program to restores retinal images. The sources and tests are hosted in *GitHub*'s repository and can be found in (David, 2016b) with *RRtoolbox*'s manual specifically in (David, 2016a). The pre-release is hosted in *GitHub* as well following the instructions in subsection 0.3 of the appendix. The *RRtoolbox* package have many potentials as demonstrated with *imrestore* and the developing *RRtoolFC* GUI explained in the Future Work section .

It is clear that the current restoration method does not check the retinal images for unwanted cases but only the quality of the image and it is responsibility of the user to provide good sets (bad case in section Result for set 6). This was not contemplated in the objectives but future work could improve upon this. Although it can be a drawback *imrestore* compensates this with robust threshold methods for more precise segmentation of objects under irregular cases (i.e. illumination conditions, blurry areas, etc.) improving on robustness, something that normal algorithms do not offer (see lens simulation results in the figures of Results section). Even more, *imrestore* replaces the use of binary masks with alpha masks to mitigate confusing areas and better control operations involving segmentations. Notice that these segmentation methods are in the color domain and they do not provide means to completely discriminate objects (i.e. the bright area of the optic disk in the retina could be treated as a normal flare). Currently the space domain is being worked on but it is still experimental (i.e. localization of custom objects like the optic disk for any case are not robust).

It was demonstrated that caching techniques improve considerably processing times (see profile tables in Results section) but can be really dangerous if data is corrupt or is not checked for its integrity causing unexpected results.

Recommendations

Though the objective is to restore set of images with defects and low resolution it is recommended that the user takes the set of images at similar lighting conditions and they should be focused preventing blurry areas. This ensures best results, remember quality images produce quality results.

When using *imrestore* it is a good practice to cache data to an absolute path using `--cachePath <my absolute custom path>` and prevent relative paths. This useful when *imrestore* is used multiple times in the same set of images to improve speed or continue from an advanced point of the program should an unexpected error crash it in a previous session.

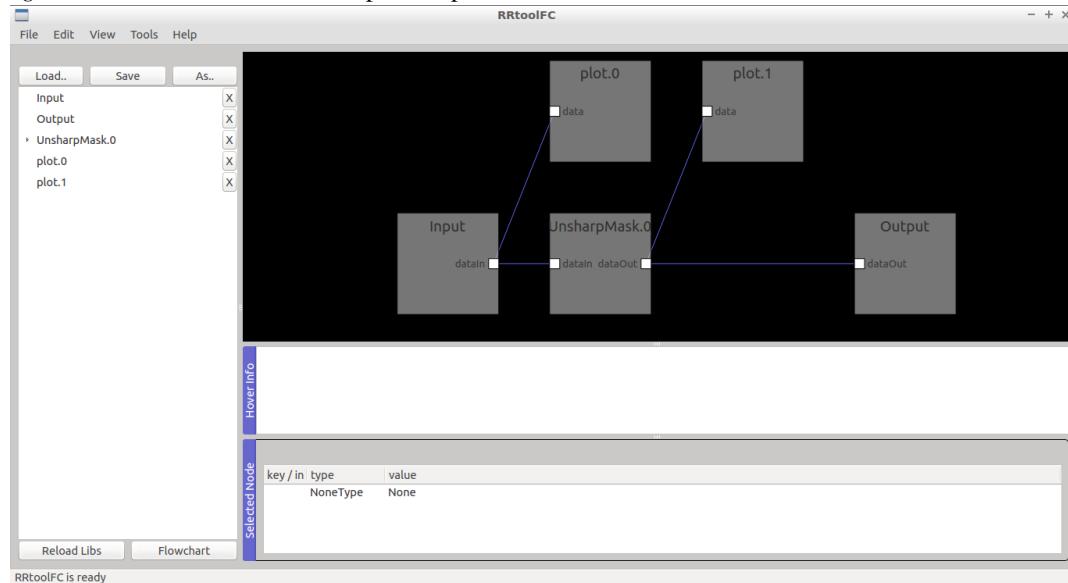
For retinal images it is recommended that only the retinal area should be in the image so that cases like in section Result for set 6 does not hinder the capabilities of the restoration. And although *imrestore* will try to change the perspective of an image so that it can fit in the restoring set it is recommended that all the images have similar perspectives (i.e. images are taken from similar positions and not too apart of the same angle).

Future Work

So much work can be done to and on top of the *RRtoolbox* package which was demonstrated with the implementation of the *imrestore* program. This is the case contemplated in the *RRtools* repository (David, 2016b). In it an additional project is being developed under the name of *RRtoolFC* which is a graphical interface designed to ease the implementation of new tools regarding the application of *RRtoolbox* for images in general or any other package that wants to be ported to it. All this with the intention of providing users with the power to develop their own useful applications.

As the *RRtoolFC* acronyms implies it is the Function Chart utility of the Retinal Restoration tools. Running its main window *RRtoolFC/GUI/main.py* loads the GUI showed in Figure 89. This interface was mainly made using *Qt4 framework* and *PyQtGraph* for the support of the fast prototyping with *nodes* (Luke Campagnola, n.d.). Sadly *PyQtGraph* is not totally mature as other programs (MathWorks, n.d.-b) and does not provide the entire services needed for *RRtoolFC* slowing the development of it. This demanded the study of the *PyQtGraph* package to learn from its development and extend it to our needs. Some functionalities are being developed to support the transformation of any python code to generic *nodes* (the blocks showed in the canvas are of Figure 89) allowing the user to port their graphical program to and from pure python code. Additionally, though the nodes can be inherited from code it cannot be accessed from the graphical interface, so it is being developed a way to allow them to be accessed and modified if necessary. This obviously has the problem of how deep internal nodes of the current node can be accessed and modified or how much information can be included when saving programs using the *RRtoolFC* application. Furthermore, it is planned to add support for installations so that the user share and use their program as standalone applications.

Figure 89. *RRtoolFC* Main developed Graphical Interface



There is a console which opens the environment where the application runs. This console can be opened in 'View -> Console' (Figure 90a). In here we can make use of the *node* instances created in the graphical interface and even control internal variables not provided by it so that the advanced user can have more control over them or debug their graphical implementation should a difficult error appear to solve them. For example Figure 90b shows how to set and input and visualize an output in the console. Typing `locals().keys()` and pressing the Enter key evaluates the expression revealing the console local variables which correspond to what is in the graphical interface. "fc" is the flowchart and "loadcv" an image loader so evaluating `fc.setInput(dataIn=loadcv(r'../../tests/im1_1.jpg', flags=0, shape=(300, 300)))` updates the flowchart and produce two windows corresponding to the nodes "plot.0" (Figure 91a) and "plot.1" (Figure 91b) placed at each step to visualize data. To show the output the *fastplt* function from the *RRtoolbox* package can be imported this is done evaluating `from RRtoolbox.lib.plotter import fastplt` and if `locals().keys()` is evaluated again it can be confirmed that `locals()` was updated and now we can use *fastplt*. Lastly evaluating `fastplt(fc.output()["dataOut"])` shows the flowchart's output (Figure 91c).

Figure 90. RRtoolFC's console

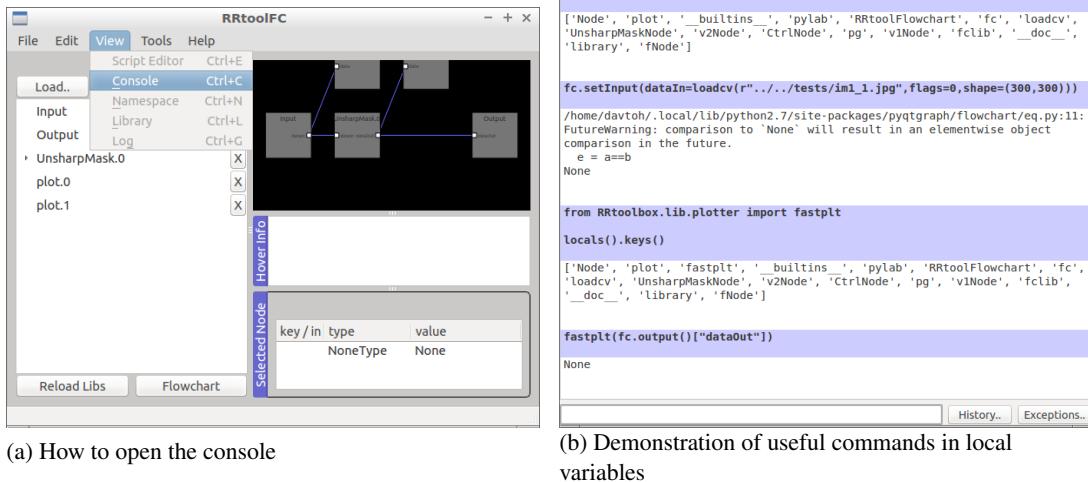
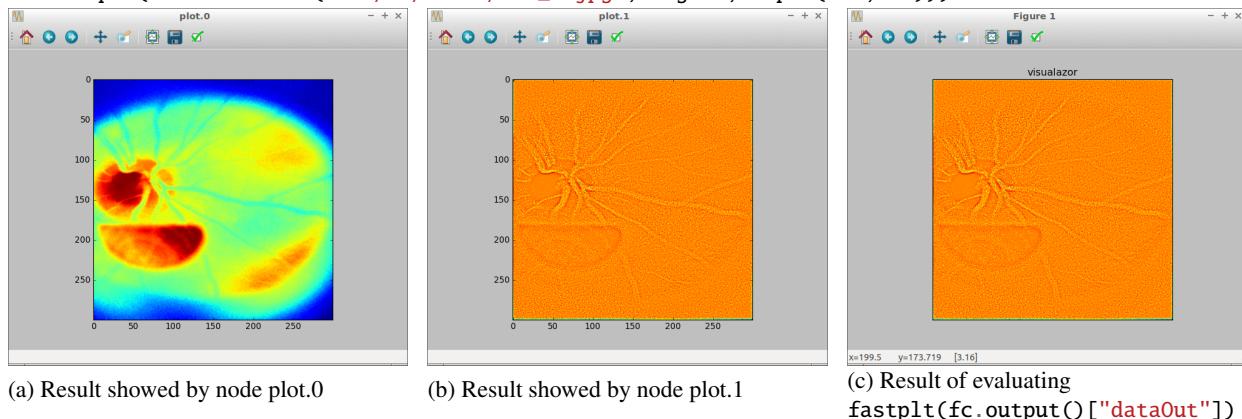


Figure 91. Functional application of the RRtoolFC's console when evaluating
`fc.setInput(dataIn=loadcv(r'.../tests/im1_1.jpg", flags=0, shape=(300, 300)))`



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Appendix A RRtoolbox: Python Implementation

Additional to the main sources of the *RRtoolbox* used to create *imrestore* it contains other algorithms used for testing, prototyping and demonstrating restoration tools. Some of the algorithm are explained here, how to set them up, their class diagrams and other program-related topics that are not in the main document.

A.1 Procedures to set-up *RRtoolbox*

It was implemented in python using some of its build-in functions and others provided via modules. If the OS is Linux there is a high probability that python 2.7 is already installed with the distribution but if that is not the case it can be installed via the package manager. Here it will be assumed that the Linux is a Debian distribution and each time something is typed in the console it will be referenced with a "\$" sign.

A.1.1 Install Python. Python is a famous programming language and support for it can be easily found in the web but for the sake of completeness it will be explained how to set it up. In Linux it is pretty straightforward:

```
$ sudo apt-get install python2.7
```

And that's it, even if the Linux distribution did have Python it should have one now. For windows, it is a few clicks away as showed below:

- Go to www.python.org
- Under Downloads select Windows
- Select the latest release for Python 2
- Download according to the Windows version and computer architecture
- Once downloaded run the installer and follow the Setup steps.
- Once installed register python executable path in a Windows variable.

Register Python's path in a Windows' variable (Window 7). If for any reason the Python installation did not register its executable in a Windows variable for it to be used as a command in the cmd then this steps can be followed to solve the issue:

- Go to Control Panel (menu)
- System and security (icon)
- System (icon)
- Advanced system settings (left menu)
- Environment variables (button)
- Under the path variable append the python variable
 - Double click in Path
 - Once the "Edit System Variable" Dialog is opened
 - Under "Variable value" box: at the end of all its variables insert the following
; C:\Python27
 - Click OK button
- Click Ok button
- Now under the command Prompt the user should be able to execute a python command.

A.1.2 Install pip. Pip is a python module used as a package manager to install more packages. It is in Linux with the package manager:

```
$ sudo apt-get install python-pip
```

And for Windows it can be installed as follows

- Go to <https://pip.pypa.io/en/stable/installing/>
 - Download the get-pip.py
 - Run the script
- ```
$ python get-pip.py
```

**A.1.3 Install packages to Python using pip.** To install a package to python using pip in Linux or Windows it just have to be typed in the console the command pip followed by the program. So, to install the principal dependencies of RRtoolbox's packages then type them one by one:

```
$ pip install numpy
$ pip install matplotlib
$ pip install dill
$ pip install joblib
$ pip install pyinstaller
```

Numpy and matplotlib packages also can be obtained installing Scipy which comes with them and other scientific modules or following the instructions at <https://www.scipy.org/install.html>:

```
$ pip install scipy
```

## Code 6: Install OpenCV

```

https://help.ubuntu.com/community/OpenCV
http://docs.opencv.org/2.4/doc/tutorials/introduction/linux_install/linux_install.html
version=$(wget -q -O - http://sourceforge.net/projects/opencvlibrary/files/opencv-unix | grep -m1 -o
 ~ '[2]([.][0-9]+)' | cut -c2-)
echo "Preparing to install OpenCV" $version
pre-requisites
mkdir OpenCV
cd OpenCV
echo "Removing any pre-installed ffmpeg and x264"
sudo apt-get -qq remove ffmpeg x264 libx264-dev
echo "Installing Dependencies"
sudo apt-get -qq install libopencv-dev build-essential checkinstall cmake pkg-config yasm libjpeg-dev
 ~ libjasper-dev libavcodec-dev libavformat-dev libswscale-dev libdc1394-22-dev libxine2-dev
 ~ libgstreamer0.10-dev libgstreamer-plugins-base0.10-dev libv4l-dev python-dev python-numpy libtbb-dev
 ~ libqt4-dev libgtk2.0-dev libfaac-dev libmp3lame-dev libopencv-amrnb-dev libopencv-amrwb-dev
 ~ libtheora-dev libvorbis-dev libxvidcore-dev x264 v4l-utils ffmpeg cmake qt5-default checkinstall
downloading
if [-f "OpenCV-$version.zip"]
then
 echo "using OpenCV-$version.zip already in folder."
else
 echo "Downloading OpenCV" $version
 wget -O OpenCV-$version.zip
 ~ http://sourceforge.net/projects/opencvlibrary/files/opencv-unix/$version/opencv-"$version".zip/download
fi
installing
echo "Installing OpenCV" $version
unzip OpenCV-$version.zip
cd opencv-$version
mkdir build
cd build
cmake -D CMAKE_BUILD_TYPE=RELEASE -D CMAKE_INSTALL_PREFIX=/usr/local -D WITH_TBB=ON -D
 ~ BUILD_NEW_PYTHON_SUPPORT=ON -D WITH_V4L=ON -D INSTALL_C_EXAMPLES=ON -D INSTALL_PYTHON_EXAMPLES=ON -D
 ~ BUILD_EXAMPLES=ON -D WITH_QT=ON -D WITH_OPENGL=ON ..
make -j2
sudo checkinstall
sudo sh -c 'echo "/usr/local/lib" > /etc/ld.so.conf.d/opencv.conf'
sudo ldconfig
echo "OpenCV" $version "ready to be used"

```

To install OpenCV 2.4.13 (or newer versions from the 2.4 release) in Linux follow this instructions:

- Create an empty file "install\_opencv.sh", open it and place Code 6.
  - Save "install\_opencv.sh" and close.
  - Run in the terminal:
- ```

$ chmod +x install_opencv.sh
$ ./install_opencv.sh

```

The script `install_opencv.sh` will download and install the latest OpenCV release of 2.4 version. It builds the release from source which takes a considerable amount of time and computer power so an stable power connection and closing all unrelated programs is recommended.

In Windows, download the installer at <https://sourceforge.net/projects/opencvlibrary/files/opencv-win/2.4.13/opencv-2.4.13.exe/> download and install.

0.2 Procedures to generate *RRtoolbox* documentation

The *sphinx* Python module is used (Georg Brandl, n.d.). The command to generate *RRtoolbox* documentation is:

```
sphinx-apidoc -F -H RRtoolbox -A davtoh -V 0.1 -R 1 -o . . ./RRtoolbox .. ./RRtoolbox/temp
```

This generates the folders *_build* used by *sphinx* to place its built documentations, *_static* to place additional sources (e.g images) and *_templates* used to place templates which modifies the look of created documents. Additionally it creates the files *Makefile*, *make.bat*, *conf.py* files used to configure *sphinx* to generate the documentation. The other *.rst* files are related to the documentation structure. To let *sphinx* find the *RRtoolbox* folder lines must be placed in the configuration file *conf.py*:

```
import sys,os
sys.path.insert(0, os.path.abspath("../")) # add RRtools path
```

Once *sphinx* is configured as desired the *make* command can be used to generate the documentation. So to generate the pdf just it has to be typed in:

```
make latexpdf
```

or to generate the html files:

```
make html
```

Additional options can be consulted asking the bash for help:

```
make -h
```

0.3 Procedures to download *imrestore* program and source

To download the execution files of *imrestore* just go to the *RRtools* repository found in (David, 2016b). Click in the link as shown at Figure 92 in the red box then when the page is loaded as in Figure 93a scroll down to the download section as in Figure 93b. Once there you can download the compiled versions for windows and Linux enclosed in the red boxes. The current source code can be downloaded too and its weight repents in the current sources, at the realization of this document the *.zip* file weights 92,2 MiB and the contents can be appreciated in Figure 94.

Figure 92. Finding release link in *RRtools* repository

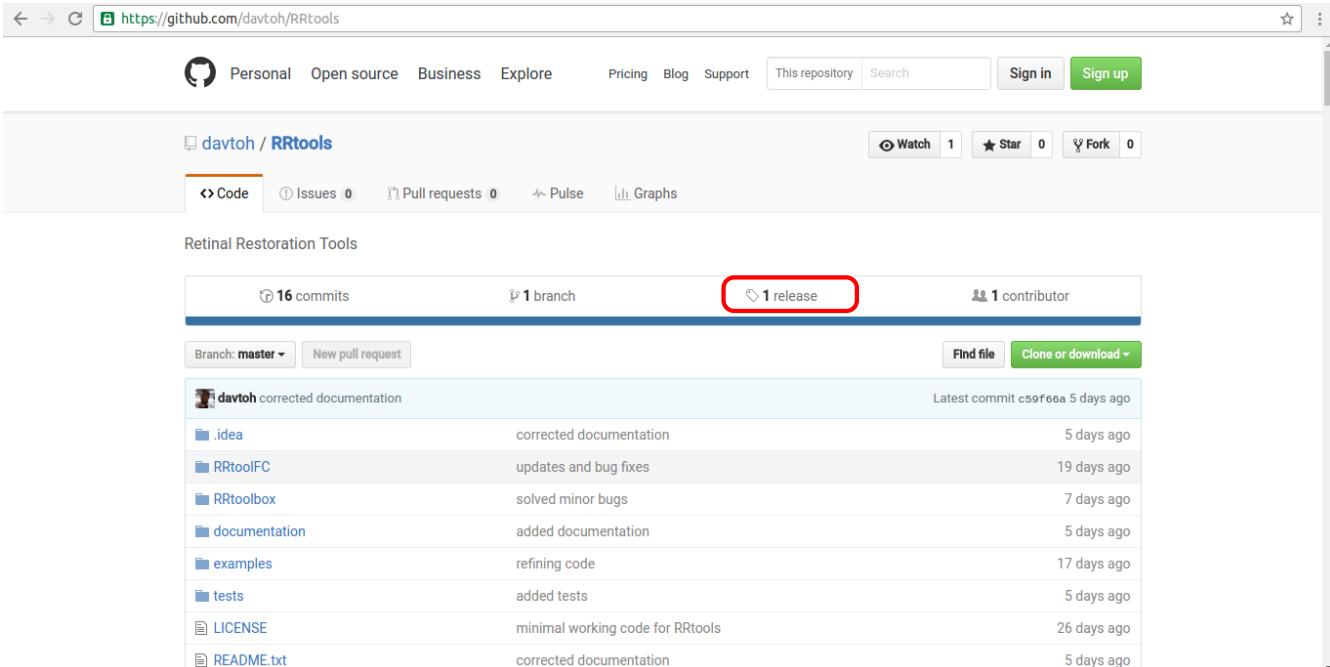


Figure 93. Imrestore pre-release

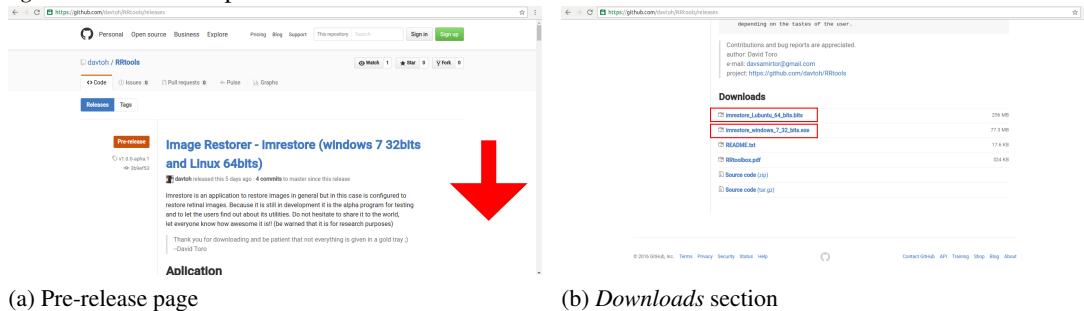


Figure 94. Imrestore downloaded sources

