

Streszczenie:

Zakresem poniższej pracy jest proces projektowania i tworzenia urządzenia, które automatycznie przypisuje rozbite jajka kurze do jednej z dwóch poniższych klas:

1. nieuszkodzone żółtka otoczone białkiem jajecznym albo czysta linia produkcyjna
2. uszkodzone żółtka albo nieuszkodzone żółtka otoczone szczątkami uszkodzonych żółtek

Implementacja urządzenia została poprzedzona badaniami, które porównują różne metody oceniania obrazu z kamery, biorąc pod uwagę ich skuteczność pod względem zużycia zasobów obliczeniowych, oraz skuteczność klasyfikacji.

Działający prototyp został opracowany i jest w trakcie testów.

Urządzenie będzie masowo produkowane jako moduł wykorzystywanych w maszynie RZ-1 firmy OVO-Tech.

Zwiększy on atrakcyjność maszyny dla następujących grup odbiorców:

1. Firmy farmaceutyczne, które produkują leki z białka jajecznego i stawiają wysokie wymagania pod względem czystości białka jajecznego
2. Cukiernie, które wymagają czystego białka jajecznego do procesów piekarniczych takich jak ubijanie piany jajecznej.
3. Firmy produkujące odżywki dla sportowców

Summary: The scope of this work is process of creating a device that automatically assigns broken eggs into one of two following classes:

1. intact yolks surrounded with egg white or clear processing plant line
2. damaged yolk or intact yolk surrounded with damaged yolk parts

The implementation of such device is preceded with research that compares various methods of recognizing the camera image, taking into account their performance in terms of computation resources, and in terms of correctness of the recognition.

The working prototype is successfully developed and tested.

The device will be mass produced as a module of OVO-TECH rz-1 egg braking machine, to improve machine usefulness for following recipients:

1. Pharmacy companies that manufacture medicines out of egg-white and require high standards in terms of egg-white cleanliness
2. Confectionaries that require clean egg white for baking procedures, such as egg foaming process
3. Athletes nutrition producers

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1 Introduction

1.1 Research scope

In 2014 creation of camera based module for chicken egg processing machines was requested by OVO-TECH company (the client).

The module is supposed to increase quality of egg-mass based products made with use of client machines.

During last 4 years multiple attempts were made by client to separate whites from yolks of freshly broken eggs in a mechanic way.

Obtained solutions proved to be working good enough for some companies, while other rejected them as imperfect and declared will to buy clients devices in the future, provided that adequate improvements will be made.

The scope of this thesis is overviewing the possibilities of resolving client needs with image processing mechanisms that will optically detect state of object (batch of intermediate product) and judge whether or not it should be discarded from the processing plant.

Moreover, a process of creating a prototype device using tested method is described.

Problems such as: - choosing image preprocessing methods - finding optimal parameters for used image alternations - overfitting issues - real-device solution implementation, - technical issues concerning the device choice, setup and performance had been taken into account during the research procedure.



Figure 1.1.1: Clients egg braking machines operating over the world

Figure 1.1 Clients egg braking machines operating over the world

1.2 Host machine

The machine that will be extended with the module prototype prepared in this thesis is OVO-TECH RZ-1 model.

Applications	Egg breaker used to produce fresh liquid egg. Enable to separate white from yolks.
How it works	The machine imitates work of people. Egg shells are cut and opened with use of special knives. Hygienic standard of broken eggs is very high.
Destination	Bakeries, confectioneries, egg processing companies.
Capacity	max 3600 eggs/h (separating function off) 1800 eggs/h (separating function on)
Power input	0.37 kW, 1 × 230 V, 50~60 Hz
Dimensions	900 × 1600 × 1100(h) mm
Minimal workspace	2000 × 1500 mm
Weight	ca. 140 kg
Operated by	1–3 people
Delivery time	5 weeks
Note	The machine includes quality control module disallowing weak yolks get into final product. The machine is equipped with yolks-whites separator, and also has 2 modes: fast and precisely.

Figure 1.2.1: RZ-1 unit basic parameters

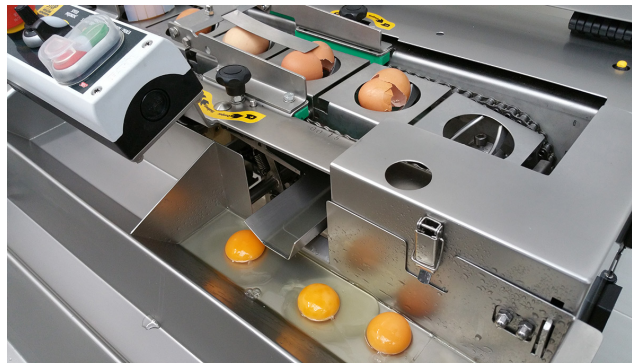


Figure 1.2.2: Egg cracking module of RZ-1 unit

The machine operates in a following way:

1. Eggs are directed into one-line flow, and mechanically aligned
2. Cracking module cuts the incoming egg surface from below with two knives that are aligned in direction of egg focal radius
3. The knives imitate human hands work by opening the previously notched egg
4. The cracked eggs are accumulated in a movable utensil for optical assessment. If the egg yolk appears to be broken, it is removed by an operator from the processing plant
5. Cracked eggs are transported by sliding to white and yolk separating module
6. Separation is done in a mechanic way while eggs slide over specifically shaped gap that only egg white fills in, while the egg yolks remain intact

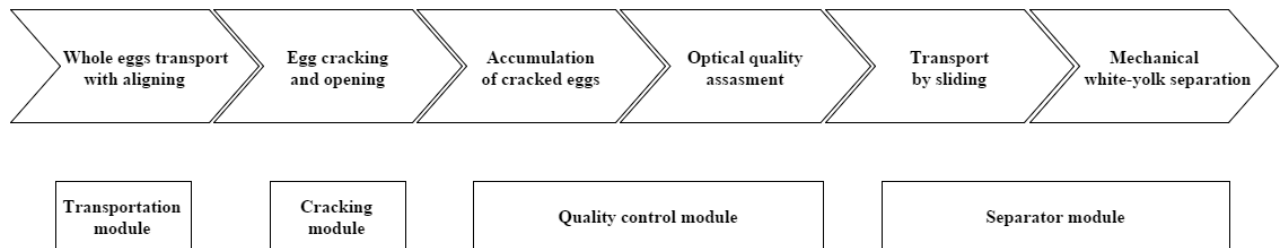


Figure 1.2.3: RZ-1 Egg processing - process and plant structure

RZ-1 operates in two modes: - Precision mode - Fast mode

Precision mode is used widely with bad quality eggs to extend amount of time for optical assessment in (5) phase.

More advanced OVO-TECH models such us RZ-3, RZ-6 and RZ-8 will be also considered for further research.

1.3 Problem description

If an egg processed by rz-1 machine is old or the chicken was feed improperly the yolk might be either broken already inside the egg, or might break during the cracking phase due to cracker imperfections such us being calibrated for different size or weight eggs.

A mixture of damaged yolk and white is created in a result, and itâ€™s not possible to separate it into two initial components using the rz-1 separation module.

Batch of such mixture should be removed from the processing plant, since it would contaminate pure egg white product otherwise.



Figure 1.3.1: Properly cracked egg

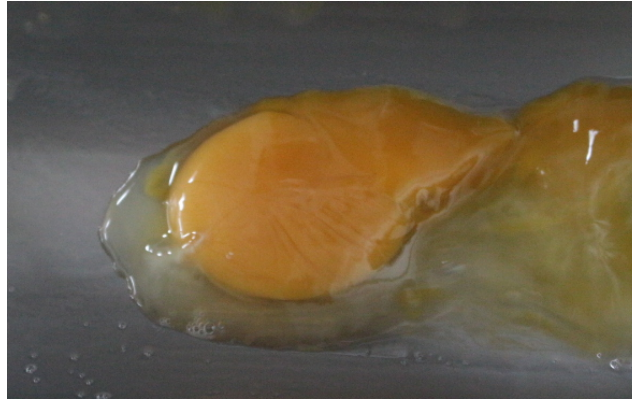


Figure 1.3.2: Damaged yolk in improperly cracked egg

1.4 Demand for improvements

Nine companies were inquired whether or not would they invest in improved Ovo-tech machine, provided that egg white will be visually clean of yolk parts.

Seven of them replied positively that they would seriously consider such offer since their processing plants would benefit in terms of product quality or manpower cost on such improvement. Remaining 2 companies did not answer the question.

One should consider why better separation is a requirement for those companies.

Following reasons were presented:

- Confectionaries and bakeries utilize egg white foam as an ingredient for their products.

Egg white foam is obtained by aeration (also known as beating / foaming / whipping) process. Comprehensive process description was provided by one of the bakers:

"When air is incorporated into a liquid or viscous solution, the solution entraps the air bubbles, forming a foam. If the foam is stabilized by proteins, it leavens a food, increasing its height and reducing its density. The viscosity of all egg products is ideal for incorporating air cells during the whipping or beating process. As whipping or beating progresses, air bubbles decrease in size and increase in number, all the time surrounded by egg proteins. Liquid egg products have low air-liquid interfacial tension; thus, when eggs are beaten or whipped, the proteins denature, or simply, they unfold. This exposes two oppositely charged ends of the protein molecule: the hydrophobic, or water-hating end, and the hydrophilic, or water-loving ends. The proteins align themselves between the air and water, securing the air bubbles with their hydrophilic chains pointing into the water and dangling their hydrophobic chains in the air. During baking, these proteins bond with each other, forming a delicate, yet reinforced network.

Egg whites do this much better than yolks because of the unique proteins found in whites. In fact, even though the term foam technically refers to any system where there are entrapped air bubbles, in the food industry, when discussing egg white products, the term tends to be exclusive to egg whites foams. This is because egg whites, unlike any other natural food ingredient, are able to create the largest possible food foam, a foam six to eight times greater in volume than unwhipped, non-aerated liquid egg white

Whole eggs and egg yolks can also increase the volume of foods, including certain baked goods and dairy desserts such as ice cream and custard, but just not as much as egg whites alone. Visually, whipped yolks may double or triple in volume, while whipped whole eggs produce less volume than either yolks or whites whipped separately. They are also less thick than yolks alone." [1]

- Ovo-tech observed that a simple dependence exists: the older the egg is, the more likely

its yolk is degenerated inside the egg shell. Thus eliminating the batches containing fuzzy yolk should decrease chances of accidentally processing the contaminated eggs, while the risk of egg contamination generally increases with egg age.

- One of the Ovo-tech clients is a biotechnology company that manufacture medicine out of genetically modified eggs.

The company purchased some ovo-tech machines, but is nevertheless displeased with their separation ratio.

This company currently is undergoing a procedure of obtaining US Food Drug Administration (FDA) approval for wide distribution of their egg-based medicine.

Executives of this client suspect that using machines with better egg separation will increase their chance for positive outcome of above process.

A non-disclosure agreement was signed between this client and ovo-tech, thus further expanding this topic in scope of this thesis is not possible.

2 Problem analysis

2.1 Establishing framework for product cleanness assessment

The requirement is to create a system able to detect all badly-cracked eggs.

After request for clarification, client made 60+ pictures of product during machine operation and classified them manually, thus requirement is less ambiguous.

Client decided that intact egg yolks should be removed from processing line if:

1. intact yolk is surrounded by homogeneous mixture of another damaged yolk and white.
This is the most strict (in terms of cleanness) case, that generates less strict cases:
2. intact yolk surrounded by damaged yolk blobs
3. fuzzy (damaged) yolk that still keeps its shape without mixing with white
4. damaged egg blobs surrounded with white
5. egg-white homogeneous mixture
6. egg-white nonhomogeneous mixture

Two cases are accepted as proper product:

7. intact yolk surrounded with egg white
8. clean egg white also, there is no need to do anything when:
9. transportation part is empty (machine is not operating or the operators are preparing next batch of eggs).

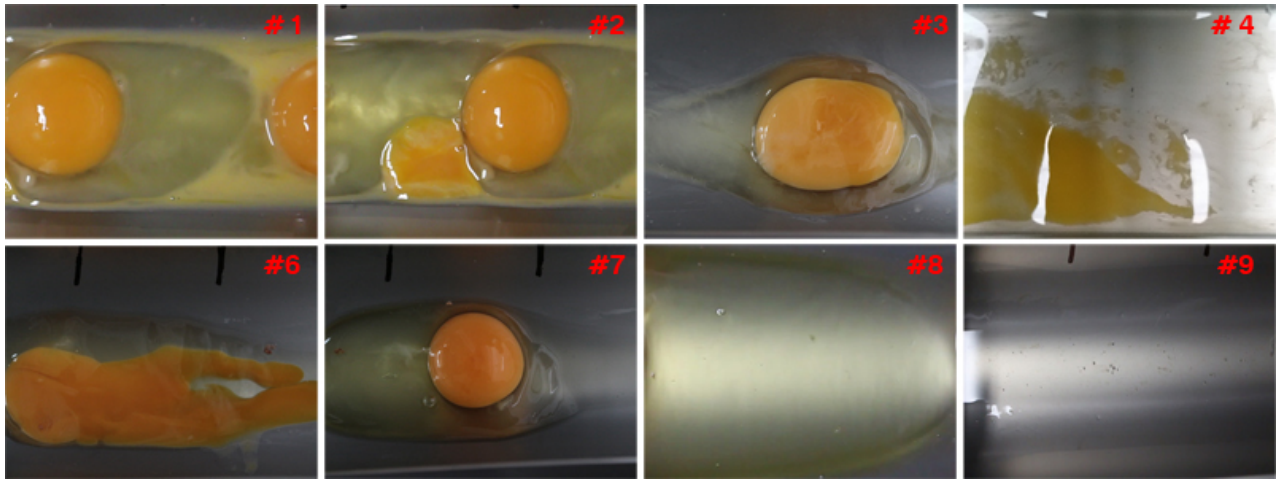


Figure 2.1.1: Cracked egg product cases. Source: own materials

Source: own materials

Also, a problem is constituted by chalaza - a usually white element that is attached to egg yolk for in-egg suspension.

Despite the fact, that client decided that chalaza position and amount are irrelevant to him, its appearance may increase difficulty of the recognition problem if identified as parts of damaged egg yolk.



Figure 2.1.2: Chalaza attached to egg. Source: own materials

2.2 Techniques of automatic egg quality assessment

Different techniques for automatic, non-destructive egg quality grading have been investigated. Among them, the following are worth mentioning:

- Intelligent systems based on visible-infrared transmittance spectroscopy[2]
- Fourier frequency analysis of a vibration-based response on impact force[3]
- Ultrasonic, magnetic resonance, electric resistance, electric nose[4]

All methods listed above require very specific equipment, are prone to overfitting and take in to account assumption that does not has to be satisfied for problem considered in this thesis: the egg shell should not be damaged.

Since in this very special case, eggs are already broken, simplified and different methods can be considered. Initially, two ideas were researched by author of this thesis:

- Proximity sensor utilization
- Using laser / narrow light beam

Both methods would identify the intact yolk basing on its height and wouldn't require much processing (an Arduino Uno or other ATmega based circuits might be sufficient).

Nevertheless, both of them had to be rejected, since intact yolks presence in the tested batch is not sufficient for batch accepting – absence of egg yolk parts is also required (see subchapter 2.1).

Finally, using images of the batch obtained from camera and processing it was chosen as a most promising option for further research.

3 Batch state detection system design

3.1 System structure

The system is expected to replace Quality control module (See. Figure 1.2.3).

The egg accumulation will no longer be required, as well as the work of qualified expert that optically assessed the egg quality is not needed anymore.

A camera will be placed on top of Separator module part, where cracked eggs are transported by sliding to the separating part.

The processing device will use the frames of camera image. Some of the frames will be skipped to reduce amount of data.

Frameskipping ratio depends on decision algorithm and device performance.

The processing device will use the frames, to make a decision whether or not the egg batch visible on the frame is acceptable, or represent a badly cracked egg.

A signal will be sent to element that will either remove egg batch from the processing plant or let it pass further.

Development of this physical element is done by OVO-TECH company and is not analysed in the thesis. OVO-TECH declared, that having +5V signal for positive batch, and -5V for negative is enough for them to work on their part.

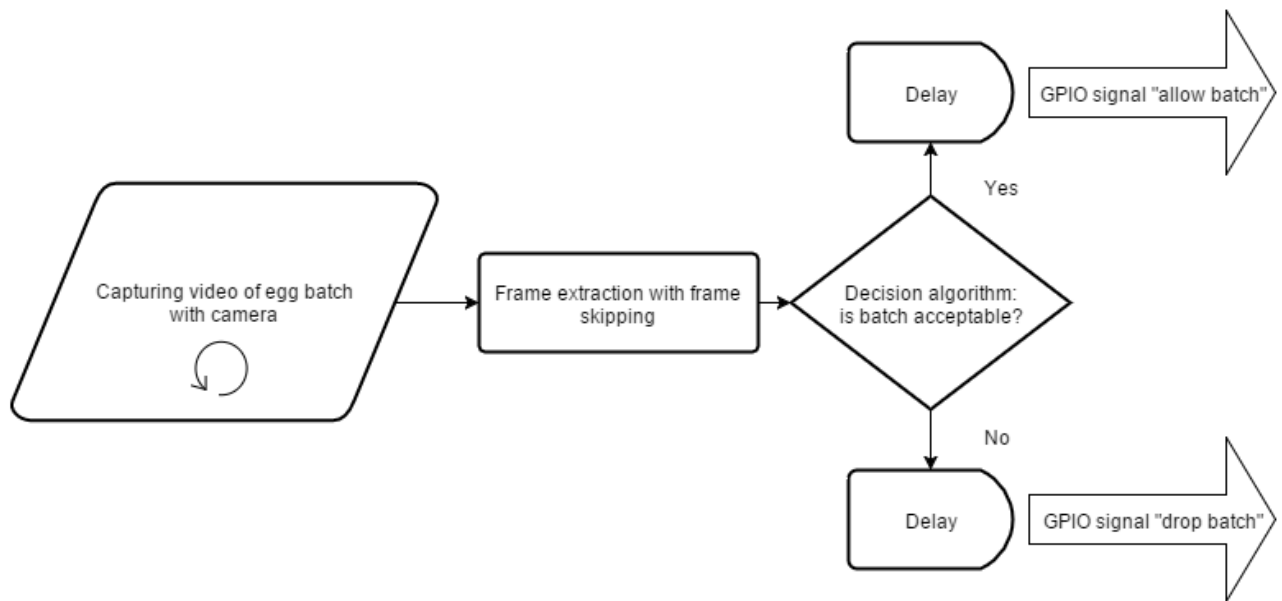


Figure 3.1.1: System structure

3.2 Detecting batch state with camera image

When it comes to assessment of camera image, 4 approaches are widely considered:

1. Very easy scanning with photo sensors or very low resolution cameras
2. Computer graphics preprocessing and segmentation accompanied with assessment of a parameter (number of pixels, number of feature points, percentage of one region that fills another bigger region and similar)
3. Machine learning approach
4. Neural networks which are used as a special case of either machine learning process or a part of it

First case was considered, since it could be implemented on Arduino Uno, other ATmega based platform or real-time systems working on PLC controllers such as LOGO! 7, which are often used in factories due to their reliability (apart from Stuxnet-class worms, that has been firstly observed in 2010, no other security threads are believed to exist).

Unfortunately, the problem is more complex than simply deciding whether the product is present or not; or whether the product is yellow or not, thus more advanced processing is required.

Second case will be expanded in the next subchapter and has been chosen as main method that will be implemented and tested in this thesis.

Last two approaches will be tested if the 2nd approach will provide to be insufficient. It is worth mentioning, that for (3) Support Vector Machines was considered, since it classifies data to exactly two classes (i.e. proper and improper egg mass).

Also, Haar Features Cascade was tested, but data amount obtained for testing was insufficient, and the methods popular implementations are focused on locating the particular object on image rather than on comparing two pattern cases.

(4) Case was considered and the author build a multilayered perceptron network working with MNIST database in order to better understand the method.
The method application will be researched further on beyond the scope of this work.

3.3 Detection algorithm

Frames are acquired from the camera module.

Than, two copies of the frame are processed independently.

Firstly, one channel is extracted from the image.

Its state is passed further as image, along all the steps.

Secondly, the channel is a subject to preprocessing, that eliminates the unimportant data and reduces the data dimensionality.
Amount of noise is reduced.

Thirdly, the image is segmented, and the shape recognition methods are used to detect existence, position and size of the egg yolk.

Independently, other copy of the image is a subject to different color scale.

Then, only one channel is extracted from the image

Finally, the information from first copy is used to remove intact egg region from second copy.
The first copy is no longer needed and its resources are freed.

Chosen channel of transformed second copy (with intact egg excluded) serve as the only image now.

It is processed in a following way:

Firstly, an attempt is made to filter out the light reflections and flares from the image.

Hopefully, during the further development, above artifacts can be removed in a physical way, by mounting and lighting the device in a very specific way.

For now nevertheless, this step is necessary.

Secondly, image is segmented with thresholding.

Thirdly, noise (defined as small blobs) is deleted.

Finally, the obtained image is expected to contain only unwanted, damaged egg mass.
Amount of pixels containing it is counted.

Decision is made basing on the obtained amount.

It takes a form of signal sent to physical element that either removes the egg batch or let it through.

The best obtained combination of preprocessing, segmentation and grading parameters will be used in a final algorithm that will be implemented in Raspberry Pi 2 device with camera module, and then tested in the Rz-1 processing plant.

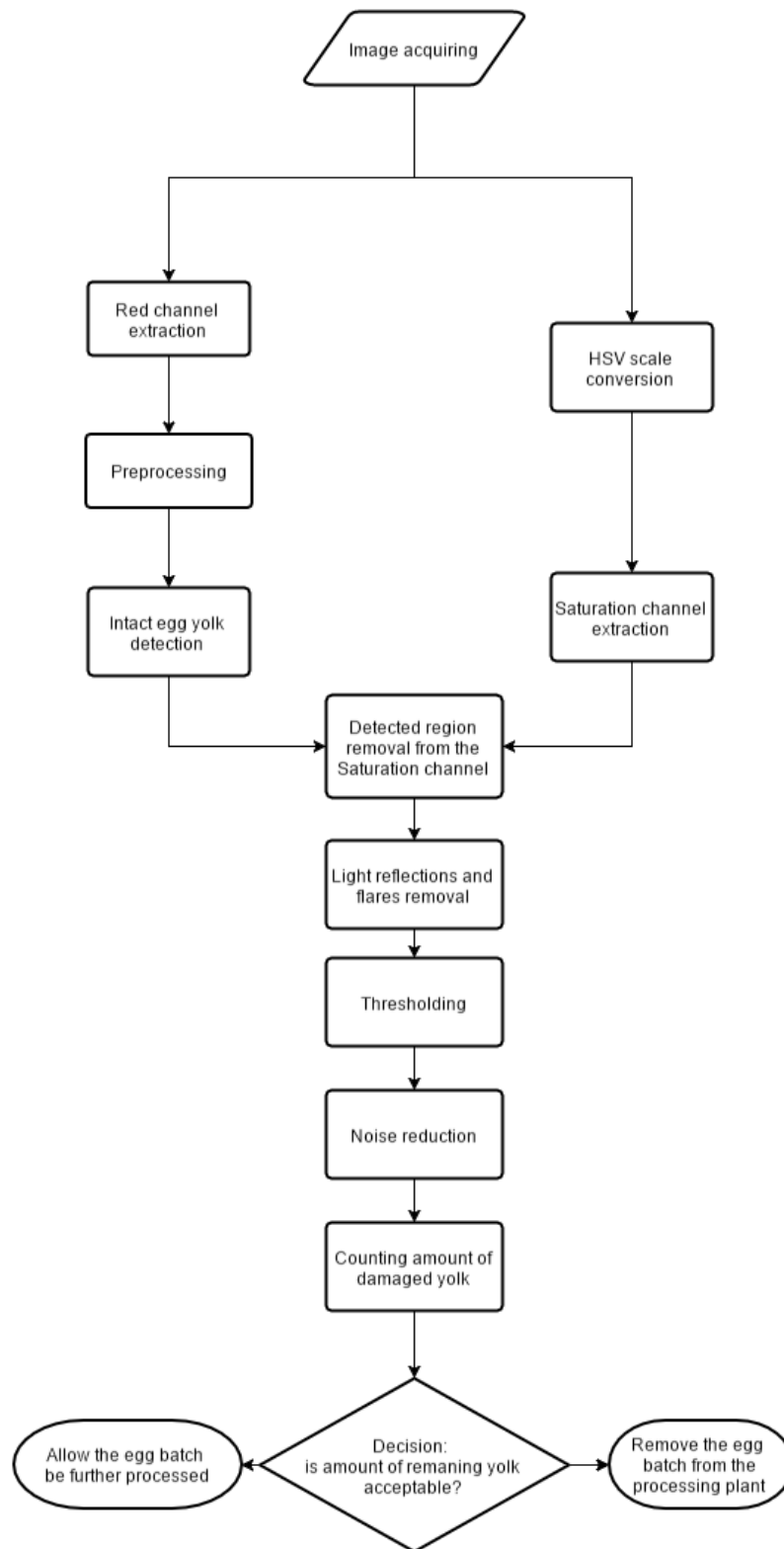


Figure 3.3.1: Detection Algorithm

3.4 Algorithm programming

Decision have been made, to utilise OpenCV computer vision library for batch state recognition.

OepnCV is designed for computational efficiency, with strong focus on real-time applications.

It has been written in optimised C and takes advantages of multicore processors. It provides simple-to-use interface that enables building sophisticated computer vision solutions.

The library features more than 500 functions that process images, a general Machine Learning Library (including Multilayered Perceptron Networks).

It is used since 1999 by major institutions such as Stanford University and is maintained and developed by specialists from companies such as IBM, Microsoft, Intel, Sony, Siemens, and Google.[5]

First recognition attempts were done by the author of thesis using Android 4.3 running on Samsung N7100 Galaxy Note II smartphone, utilising its built-in camera.

The initial code was written with use of Java distribution of OpenCV library, as well as Android SDK.

Sadly, the performance of developed pre-prototype was very poor and a single frame was processed in more than 3 seconds.

Another version was built in a way that most resource consuming operations (such as Hough Circle Transform and all array operations) were rewritten in C language.

For that purpose original C version of OpenCV was utilised, as well as NDK (Native Development Kit for Android).

Nevertheless, author of this thesis, decided that despite Android availability for embedded platforms, it is not a platform suitable for factory tasks, due to big overhead, long booting and variety of functions that will not be utilised.

Raspbian Linux "Jessie" distribution was picked as the adequate one to host recognition algorithm accompanied with Python 2.7 language.

Python is well supported on Raspian, it produces small amount of code, and can be executed fast in terms of computational time if the specific C-written functions are called.[6]

For author, a case occurred, that 39-lines code snippet that recognise nested regions of image, written in C took only 6 code lines after rewriting it in Python.

4 Egg yolk detection

4.1 Methods overview

An egg yolk, when intact and fresh can be approximated with sphere which projects as circle-like shape on 2D image.

Methods presented below enable to filter out the unnecessary noise, simplify the image and detect the yolk shape.

While not all of them were used in the final detection algorithm, they were giving promising results on the early stage prototypes.

Following methods treat images as 1, 8, or 24-bit encoded matrices, thus both terms are used synonymously.

x and y denote pixel position.

4.2 Circle Hough Transform (CHT)

Circle Hough Transform is an algorithm used for retrieving 3 parameters defining a circle-like shape on images[7] : - (x_c, y_c) - position of shape's center - r - its radius

To understand the way CHT works, simplified case is considered:

The image is 1-bit encoded; the radius r and position (x_p, y_p) of a point lying on the searched circle are known.

Potential circles space is created by treating points on a circle of radius r centered in (x_p, y_p) as centers for potential circles.

For every potential circle, number of intersecting points is counted.

The potential circle with largest number of intersections is chosen as the best fitting one.

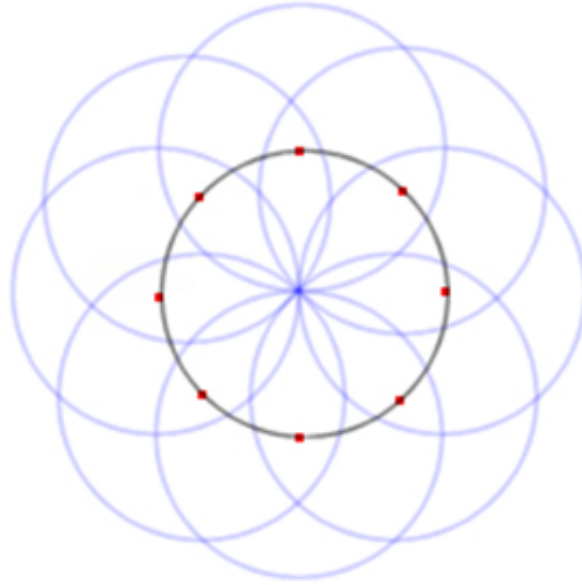


Figure 4.2.1: Potential circles space[8]

In the full cases, such procedure is repeated for every image point, and the intersections numbers are stored in accumulation matrix.[9]

To reduce computation time and false detection rate, minimum distance between image centers is given as a parameter min-dist.

Threshold value param-2 is defined as minimum number of intersections to treat circle candidate as detected circle.

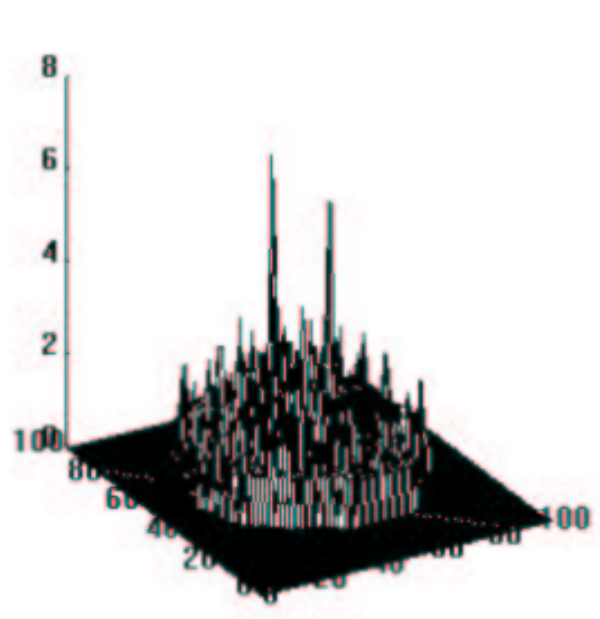


Figure 4.2.2: Accumulation matrix with two good circle candidates[9]

If r is unknown, min-radius and max-radius parameters are defined, and the algorithm repeats the procedure for all possible integer r in range of above parameters.

OpenCV utilizes Canny edge detector for finding the potential intersection points, thus operates not only on 1-bit images, but also in 8-bit grayscale and 24-bit bgr images and provides smaller number of false positives.[10]

param-1 is introduced as higher threshold for Canny edge detector, while the lower threshold is set to $0.5 * \text{param-1}$ [11] .

4.3 Erosion and Dilatation

Erosion and dilatation are two morphological filters that enable to filter out unnecessary noise from the image.

Combined, they leave the core information (biggest blobs) intact or amplified, while small elements are deleted.

It is easily applied to binary images:

A ROI (region of interest, a rectangular part of image) moves through image pixel by pixel and is compared to previously defined kernel (binary matrix, usually symmetric with respect to both diagonals).

0	0	0	1	0	0	0
0	0	1	1	1	0	0
0	1	1	1	1	1	0
1	1	1	1	1	1	1
0	1	1	1	1	1	0
0	0	1	1	1	0	0
0	0	0	1	0	0	0

Figure 4.3.1: A diamond-shaped kernel[12]

The consecutive pixels of on ROI that matches kernel elements with value 0 remain intact.

Pixels of ROI that matches kernel 1-valued elements changes to:

1-s if at least one of them is 1 for Dilatation

0-s if at least one of them is 0 for Erosion.

Erosion gives the effect of “shrinking” the objects (and if they are small enough, they are deleted).

Dilatation makes the objects bigger.



Figure 4.3.2: Original image, eroded image, dilated image.[13]

When these methods are used multiple times, and convoluted, they filter out the noise. Perimeter Determination and Skeletonization are widely used contour detection functions which utilize erosion and dilatation[13].

4.4 Gaussian blur

Gaussian blur is obtained by convoluting pixel with gauss function values.

It reduces noise, detail and filters out the high frequencies what makes it a low pass filter.[14]

Gaussian blur drastically reduces number of false-positive Hough circles detections. It nevertheless should be coupled with sharpening to reduce false-negative cases.[15]

$$G(x, y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

Figure 4.4.1: Gauss function in two dimension[16]

sigma, ksize.width, ksize.height parameters are introduced.

Ksize stands for size of precomputed kernel, that is convoluted with ROIs on the image.

The term "convolution" might be misleading and for the scope of image processing tasks is defined as picewise multiplying two matrices and than summing the obtained values:

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} * \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = (1*a)+(2*b)+(3*c)+(4*d)+(5*e)+(6*f)+(7*g)+(8*h)+(9*i).$$

Figure 4.4.2: Convolution of 3x3 matrices[17]

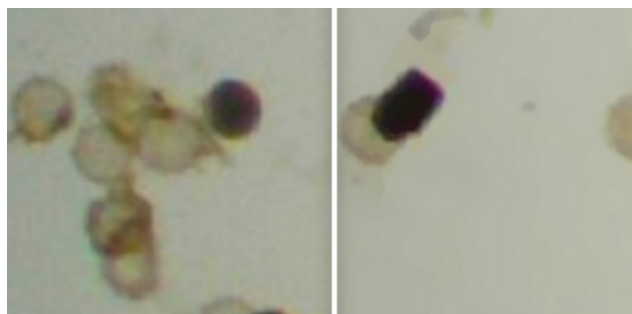


Figure 4.4.3: Microscope image before and after applying gaussian filter[15]

4.5 Channeling and HSV conversion

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4.6 Using the methods

```
def detect_yolks(frame):
    b,g,r = cv2.split(frame)
    hsv = cv2.cvtColor(frame , cv2.COLOR_BGR2HSV)
    h,s,v = cv2.split(hsv)
    g = cv2.cvtColor(frame , cv2.COLOR_BGR2GRAY)
    # ...
    # 'c' variable - channel is chosen among the above
    c = cv2.GaussianBlur(c, (kernel ,kernel), 0)
    c = cv2.threshold(c,threshold ,255,thresholding_method)
    circles = cv2.HoughCircles(c,cv2.HOUGH_GRADIENT,1 ,dist ,param1 ,
                                param2 ,minRadius ,maxRadius)

    return circles
```

5 Contaminated product recognition

5.1 General idea

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5.2 Thresholding

Thresholding is a simple case of image segmentation.

It can be used to convert one channel (in most cases grayscale) image to binary image.

The basic binary thresholding mode sets every pixel of destination matrix to:

- 0 if the source matrix value is smaller than threshold (integer parameter)
- 1 if the source matrix value is bigger than threshold.

The truncate version sets every pixel of destination matrix to

- threshold if the source matrix value is smaller than threshold, or
- leaves it unaltered otherwise.

Various modes and modifications are used depending on coding schema and colors distribution.

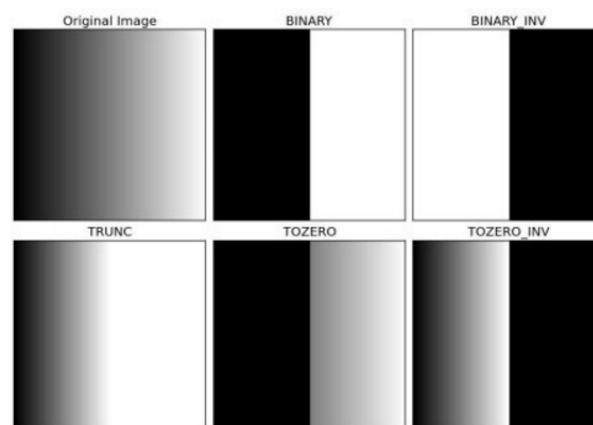


Figure 5.2.1: Different Thresholding modes generated with Metaplotlib

Finding optimal threshold value finding is a time consuming process and it often gives satisfactory results only for small set of images.

Therefore Otsu Binarization algorithm for automatically selecting threshold is used.

Moreover using one threshold value for whole image may not work properly if the lighting conditions differ in different image areas.[18]

Adaptive Thresholding utilize different threshold values for different ROIs.

Two modes are often used: Adaptive Mean and Adaptive Gaussian.

First compute threshold value as mean of neighborhood values.

Second uses weighted sum neighborhood values.

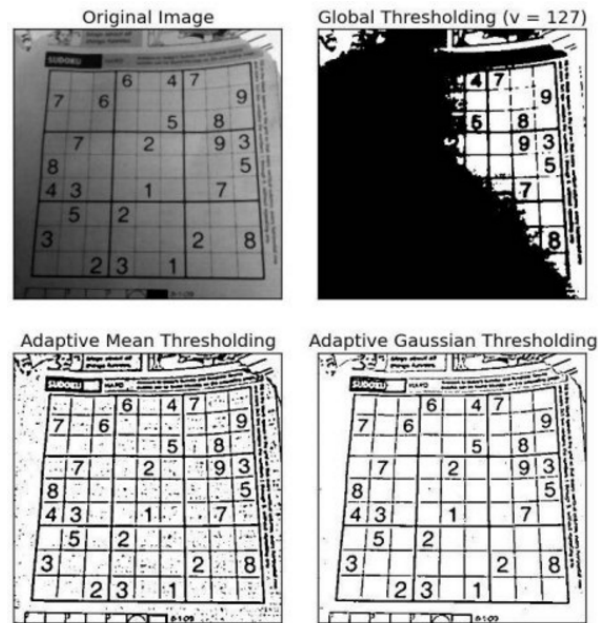


Figure 5.2.2: Threshold computing methods12[18]

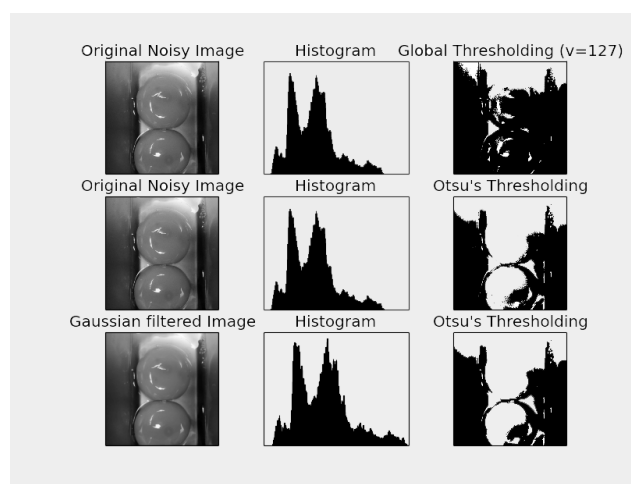


Figure 5.2.3: Different thresholding applied Fig. generated with Matplotlib

For more general cases (i.e. multichannel images, assigning more colors) clustering method known as k-means is widely used.[19]

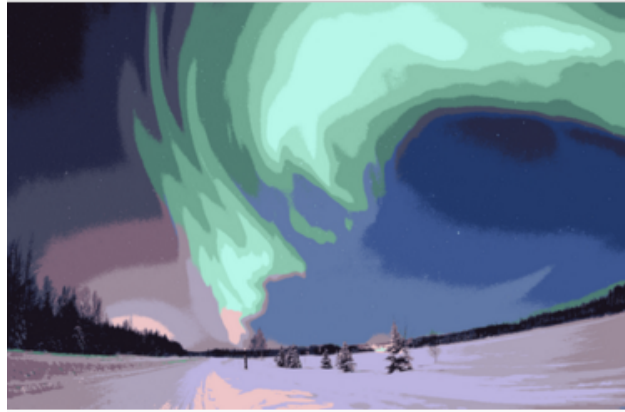


Figure 5.2.4: K-means clustering method for $k = 16$ [20]

5.3 Using the methods

```
def grade_frame(frame , circles):
    b,g,r = cv2.split(frame)
    hsv = cv2.cvtColor(frame , cv2.COLOR_BGR2HSV)
    h,s,v = cv2.split(hsv)
    g = cv2.cvtColor(frame , cv2.COLOR_BGR2GRAY)
    # ...
    # 'c' variable - channel is chosen among the above
    if isinstance(circles , np.ndarray):
        circles = np.uint16(np.around(circles))
        for i in circles[0,:]:
            cv2.rectangle(s_c ,(i[0]-i[2]-offset ,0),(i[0]+i[2]+offset , fram
            ellipse = cv2.MORPH_ELLIPSE,(kernel_size ,kernel_size)]
            kernel = cv2.getStructuringElement(ellipse)
            c = cv2.morphologyEx(c , morphology_type , kernel)
            sum = np.sum(c)
            return 0 if sum > detction_threshold else 1
```

6 Choosing detection system parameters

6.1 Manual parameters finding

There exist a set of parameters that define the mode, that previously explained methods operate as well as their instensivity.

Finding optimal values of this parameters is a hard task, since for example CHT method didn't found a single egg yolk with default numbers.

The following values have to be found:

param1 - lorem ipsum

threshold - lorem ipsum

A script was developed, that displayed over 500 pictures for 0.3 second each in a loop. Trackbars was displayed above the frames, with every trackbar corresponding to one parameter that is supposed to be chosen. Beside, the already transformed frames after transormations were shown.

The expert was reviewing the output while changing the trackbar positions to obtain an image that segmentates the egg yolk properly.

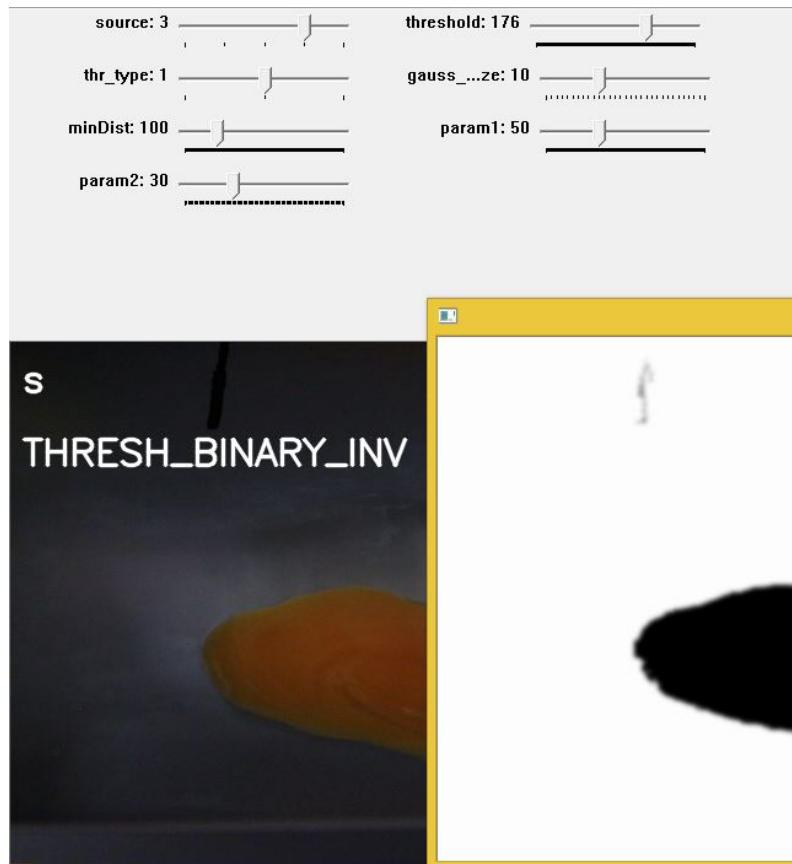


Figure 6.1.1: Screen of running one of the scripts used for manual parameter finding

Script crucial fragments:

```
cv2.createTrackbar('source', 'Egg', 3, 4, nothing)
cv2.createTrackbar('threshold', 'Egg', 176, 255, nothing)
cv2.createTrackbar('thr_type', 'Egg', 1, 2, nothing)
cv2.createTrackbar('gauss_kernel_size', 'Egg', 10, 30, nothing)
cv2.createTrackbar('minDist', 'Egg', 100, 500, nothing)
cv2.createTrackbar('param1', 'Egg', 50, 150, nothing)
cv2.createTrackbar('param2', 'Egg', 30, 100, nothing)
# ...
param2 = cv2.getTrackbarPos('param2', 'Egg')
# ...
```

Unfortunately, while some parameters sets worked fine for some image sets, they failed with another ones.

Nevertheless, the interesting intervals were obtained for each trackbar. These intervals guaranteed that at least some of the graded images are segmented in expected way.

This method resulted in excluding most parameters values and vastly reducing the optimising problem size.

6.2 Automatic parameters finding

A decision was made, to automatise process of assesing the segmentation and detection efficiency to get to optimal set of parameters.

This process was done twice:

Firstly, the egg yolk detection process was optimised.

Secondly, the contaminated product detection utilising the previously otpimised egg yolk detector was perfected.

To accomplish first objective, avaiable images were devided into 3 subsets for first process:

- 0 intact yolks visible in a frame
- 1 egg present
- 2 eggs present

For the whole set, the recognition error was defined as sum of missdetections for every subset. A simmlar procedure was conducted for second objective: the same images were devided into following subsets:

- Contaminated batches
- Batches with either only intact eggs or no egg mass at all

Error was defined as false negatives (good batch recognised as bad batch) number summed with false positives number multiplied by 3.

Multipilication by '3' weight was conducted on false positives, since keeping the product clean is more important for the client, that utilising all the egg mass.

A script was developed, that automatically loaded the images and graded them for different parameter values.

This process was very expensive in terms of processing time, and the parameters range and step was choosen to fill a night of i7-3520M 8GB RAM computer work.

It was computed in a following way: elapsed time of one full imageset testing multiplied by number of possible paramter sets.

The process was repeted by the author multiple times with different parameter intervals discovered in a process descripted in previous subsection.

Various effort were made by the author to optimise imageset grading process, so bigger variety of parameters can be checked.

Initialy, files were open prior to every image evaluation, later on all the images were loaded once, and every processing (channel splitting, colorscale conversion) was conducted before execution of the code nested in multiple loops.

Crucial parts of the testing code:

```
for param1 in xrange(10,80,5):
    for param2 in xrange(10,90,5):
        for kernel_size in xrange(1,30,2):
            for offset in (5,80,5):
                # ...
                grade_frame( detect_yolks( frame ))
```

6.3 Overfitting

Overfitting is an unwanted effect that appears often in statistics and supervised machine learning.

Overfitting means, that parameters of some model are fitting the specific modeled data case, rather than the general model.

In this case it means, that the parameters of image processing that were found, maximise the recognition efficiency for the data provided by OVO-TECH, while a new data (with different kind of eggs, different kind of noise or with images made in different conditions) can be classified in improper way.

Overfitting occurs mostly if:

- A set of sample data is too small
- A set of sample data is not diverse enough
- No regularisation or cross-validation methods are used

To reduce this effect, author used different dataset for getting the transform parameters, and different set (excluded 10

During prototype in factory testing and further development this issue will be addressed in extended way.

7 Physical system implementation

7.1 Physical Platform

7.2 Camera module

7.3 Programming raspberry pi

```
from picamera.array import PiRGBArray
from picamera import PiCamera
import time
import cv2
# initialize the camera and grab a reference to the raw camera capture
camera = PiCamera()
camera.resolution = (640, 480)
camera.framerate = 32
rawCapture = PiRGBArray(camera, size=(640, 480))

# allow the camera to warmup
time.sleep(0.1)

# capture frames from the camera
for frame in camera.capture_continuous(rawCapture, format="bgr", use_video_port=True):
    # grab the raw NumPy array representing the image, then initialize the occupied text
    image = frame.array
    # clear the stream in preparation for the next frame
    rawCapture.truncate(0)
```


8 Further research and development

Work on the solution developed in this thesis will be continued further on by the author and OVO-TECH company.

The following alternative approaches and improvements in detection system will be taken into consideration

8.1 Operating Systems

Using alternative system operating on Raspberry Pi 2 platform might improve image processing performance. Therefore, more frames per second could be analyzed, more filters can be applied and higher resolution pictures could be used. Among over 20 available ARM systems, presented below are most promising:

- Minibian - a Raspbian image that is resource optimised.
Its current Raspian "Jessie" based version boots in 14secs, use only 29MB RAM and take 451MB (compared to 1,2GB in original version).
GUI, application clients, office suite and other non-necessary packages have been removed.
The image is targeted for embedded or server applications (NAS, Web server, electronic applications).
- Arch Linux ARM
- Gentoo
- SliTaz
- Windows 10 IoT Core - embedded solution developed by Microsoft. Guarantees extensive technical support.

Largest reliability and performance benefits are expected from using realtime-capable systems.

Following solutions are available or developed for Raspberry Pi:

- RISC OS
- Xenomai
- PREEMPT_RT kernel patch and High Resolution Timers[21]
- ChibiOS/RT[22]
- RODOS - an Open Source kernel project developed by the German Aerospace Center and Prof. Montenegro's University[23]

Nevertheless advanced image processing, GPU access, utilisation of existing graphics libraries might be not available or require amount of effort unproportional to those benefits. Also, external real-time clock (such as AD9850 Pulse generator) might be required.

8.2 Platforms

- Arduino Uno R3 accompanied with DS1307 Real Time Clock - considered, but for now the image processing is too extensive in terms of image processing.
Having extremely cheap clones (4-20\$ on retail market) that feature often stronger ATmega units is an advantage although.
- Intel Galileo Gen 2 - based on Intel Quark 32-bit SoC X1000 processor, provides pins-compatible with shields designed for the Arduino Uno R3, what makes communication with other OVO-TECH machine modules easier and allows use of the mentioned above RTC.
- PandaBoard - the OMAP4430 features 1 GHz dual-core ARM Cortex-A9 MPCore CPU (a big improvement compared to Cortex-A7 CPU in Raspberry Pi 2), 304 MHz GPU, and 1 GiB DDR2 SDRAM.
What is interesting, it provides a real-time clock, what makes it a candidate for real-time system (see section 8.1).
- BeagleBoard xM - based on dual, 1.5GHz clocked ARM Cortex-A15 + dual ARM M4, 532 MHz GPU, often used with RISC-OS, having cheaper clones with more RAM.
It is considered a stronger Raspberry Pi alternative.
- Nvidia Jetson TK1 - build on quad ARM Cortex-A15, NVIDIA Kepler 192-cores CUDA GPU, 2 GiB RAM, optimised for OpenGL and OpenCV, equipped with SATA connector.
TK1 is the strongest available unit, what makes it also the most expensive one.

8.3 Server and logging

Gathering the data about egg quality, recognition performance and effectiveness is a possibility that has to be addressed.

Raspberry Pi 2 platform is capable of working as a server and thus provide system manufacturer a remote access to its operations.

8.4 One month prototype testing and environment of work tuning

The prototype will be attached to the rz-1 machine operating at OVO-TECH headquarters for a month.

During that time, led lights will indicate proper and improper egg batch.

Recognition efficiency is now > 90%. Two cases may occur, when the device is operating in working environment:

- Recognising efficiency will rise dramatically, since the camera position will be fixed, and the background (separation module slope) will have uniform texture and color.
The data provided by OVO-TECH for prototype construction covered various situations and backgrounds that will be invariant in factory application.
- Recognising efficiency will drop, due to varying egg quality, color and shape.

Two things will be done to improve image material quality registered by camera:

- Polarizing filter will be attached to the camera.
It should eliminate or drastically decrease amount of white light reflexes.
Those reflexes that make yolk appear as non-yolk at segmented image.

- Lighting will be uniform due to covering camera and a fragment of transportation slope in a box-like container with invariant, non-shadowing lights inside of it.

8.5 Testing new approaches

Current results (over 90

Two approaches have came up after the prototype development and are to be implemented and tested:

- Circularity testing - height of the segmented yolk flow can be treated as a circle diameter. Than, area of obtained circle can be compared to area of observed egg yolk. If the difference is significant, it indicates that observed blob is not a whole egg. Using a derived method for positioning the center of such circle might work better than methods used up to now if a noncomplete part of egg is visible at the image.
- Watershed algorithm accompanied with extracting peaks of Euclidean Distance Transform is belived to work better than contour detection and thresholding methods with touching or overlapping objects. It might provide efficiency boost in case of RZ-1 processing plant operating in modes faster than precise mode.[24]

8.6 Obtaining more image data and parameters tuning

Larger dataset of more different egg types will enable to better tune the processing parameters. Since there are more than 10 parameters to adjust and the number of operations that are required rise exponentially with parameters number and is multiplied by sample amount, optimising such multivariate function is a hard task.

Genetic algorithms are known to be a good metaheuristics for such tasks and should give better results in much less computation time that brute force parameter testing.

Author developed a program that implements genetic algorithm in Java, that optimises 1,2 and 3-variate functions to understand and learn the genetic algorithm concepts.

Further on, a ready, popular library will be used for this purpose: a DEAP genetic programming library for Python.



Figure 8.6.1: Screen of Java genetic function optimiser written by author of this thesis [12]

8.7 Machine learning

9 Conclusions

All the Web. sources have been checked for availability in November 2015.

Web links are not included if articles are to be immediately found with search engines, which is consistent to MLN referencing convention.

For modified web documents, date of last modification is shown.

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