

# Image processing for extracting features

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# Rockfalls

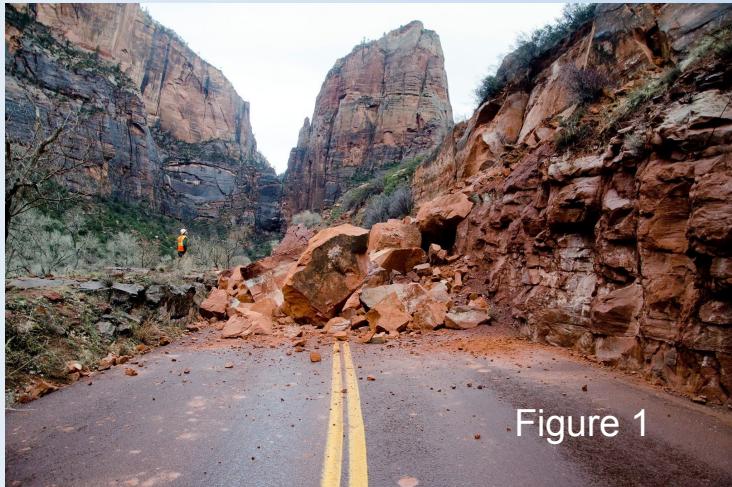


Figure 1



Figure 2

# Experimental setup



Figure 3



Figure 4



Figure 5

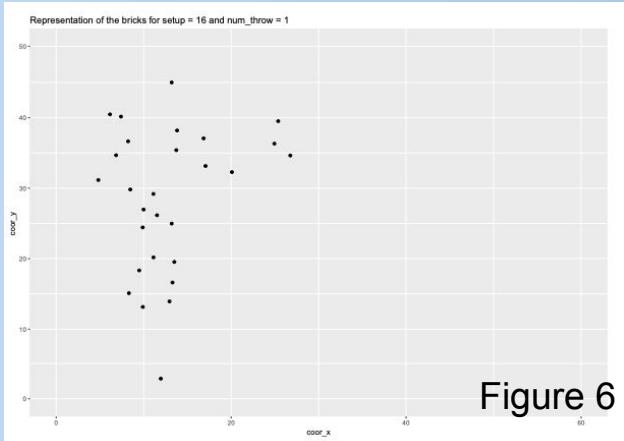


Figure 6

## **Goal :**

- Find the position of the blocks in the tray
- Result on the same rectangular pattern

## **Strategy:**

1. Tray extraction and homography
2. Block extraction inside the tray
3. Display the result

# Plan

1. Tray extraction & homography
2. Blocks extraction
3. Application
4. Results & recommendations

# Tray extraction

Tray extraction algorithm consists of two main parts:

- finding coordinates of the tray 4 corners
- homography

We are going to talk about both of it consistently.

# Tray localization algorithm overview

Main steps of the tray corners finding algorithm are:

1. Edge detection
2. Finding long enough line segments
3. Finding points of intersection

Let's consider each step more precisely using the example image.



Figure 7

# Edge detection using Canny algorithm

Canny edge detection<sup>[1]</sup> consists:

1. Smoothing (to suppress noise)
2. Finding gradients intensities and directions
3. Extraction candidates to be edges and its adjustment

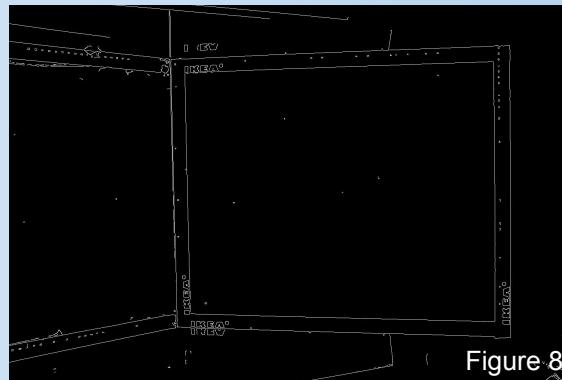


Figure 8

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[1] Canny, J., A Computational Approach To Edge Detection, IEEE Transactions on Pattern Analysis and Machine Intelligence, 8(6):679–698, 1986.

# Finding long enough line segments

Special type of Hough transform<sup>[2]</sup> - probabilistic Hough transform<sup>[3]</sup> allows us to extract the most probable line segments longer than chosen minimal length.



Figure 9

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[2] Hough, P.V.C. Method and means for recognizing complex patterns. U.S. Patent 3,069 654, December 18, 1962

[3] N. Kiryati, Y. Eldar, A.M. Bruckstein, A probabilistic Hough transform, Pattern Recognition, 1991, [https://doi.org/10.1016/0031-3203\(91\)90073-E](https://doi.org/10.1016/0031-3203(91)90073-E).

# Binarization and skeletonization

We have already extracted all necessary information from image and now we want to work only with obtained line segments, what is why we do such binarization: we drawn each pixel corresponding to found on previous step line segments with 1 on the image filled with 0.

Skeletonization is reducing binary objects to 1 pixel wide representations without losing connectivity.

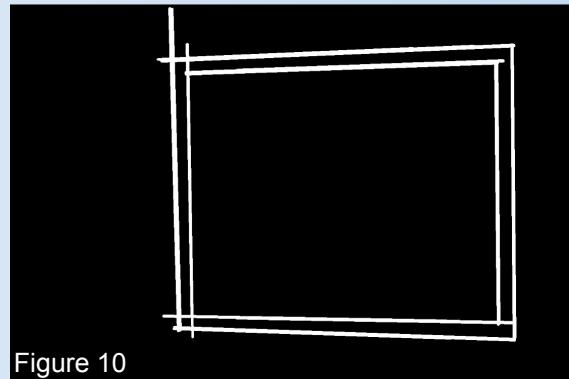


Figure 10

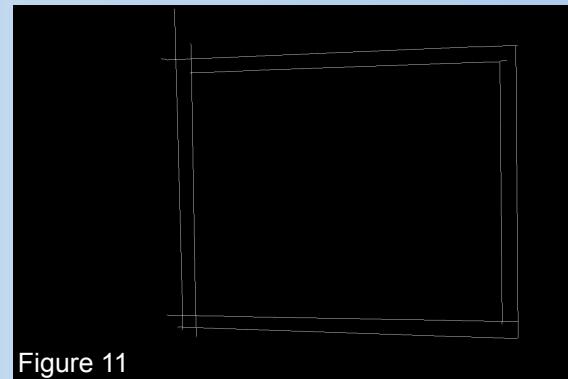
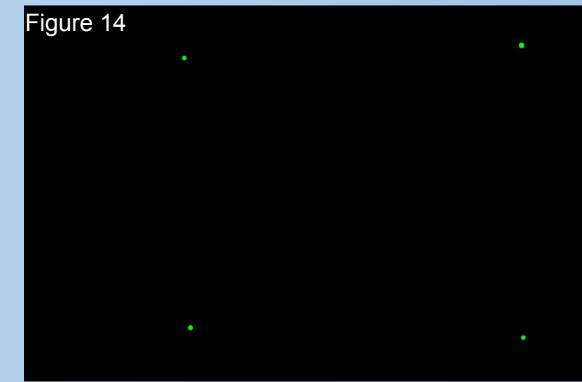
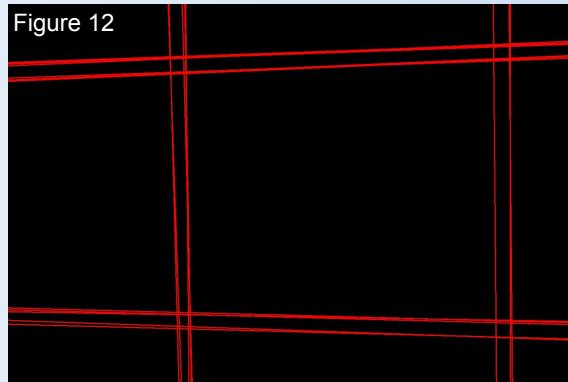


Figure 11

# Corners finding

Remaining part of the algorithm consists only in:

1. Finding straight lines going through obtained thin lines related to inner and outer sides of white tape
2. Finding its points of intersection
3. Extracting from the obtained set of points correct points corresponding to the corners



## Good cases

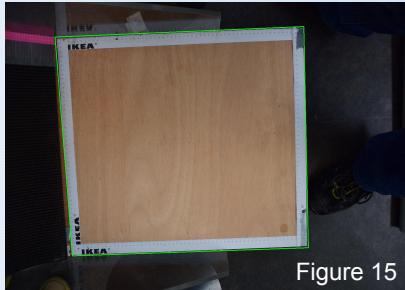


Figure 15

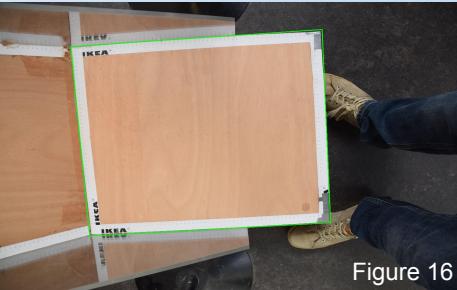


Figure 16



Figure 17

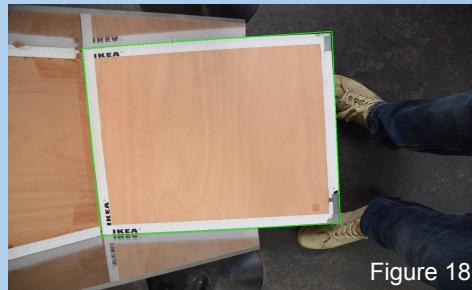


Figure 18

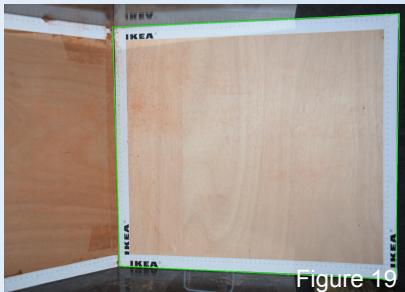


Figure 19

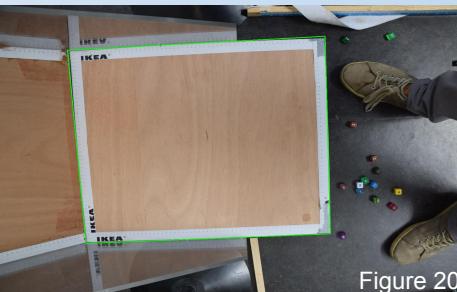


Figure 20

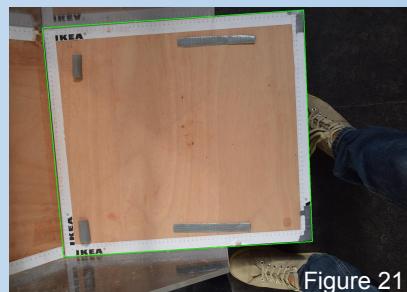


Figure 21



Figure 22

## Bad cases

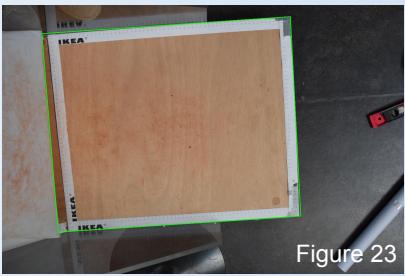


Figure 23



Figure 24

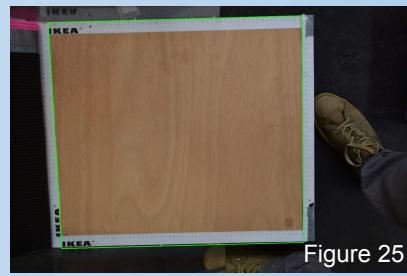
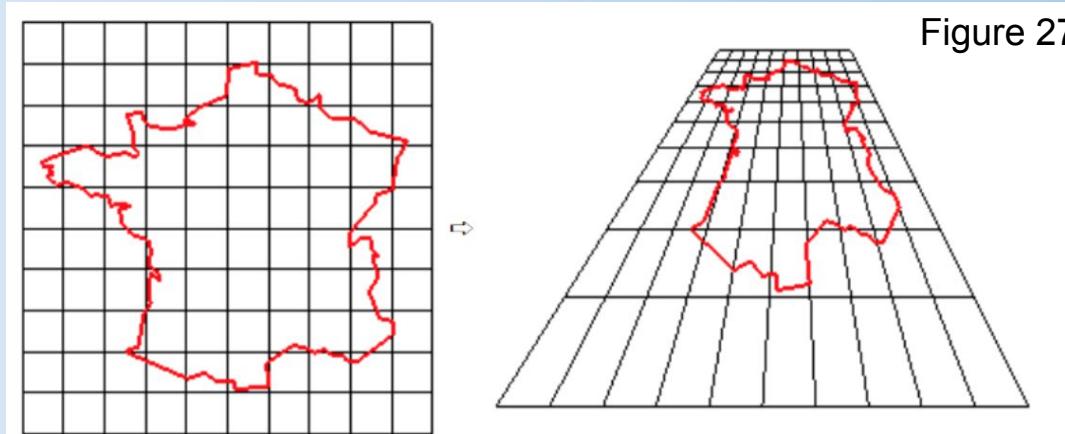


Figure 25



Figure 26

# Homography



$$(x, y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

homogeneous image coordinates

Converting *from* homogeneous coordinates

$$\begin{bmatrix} x \\ y \\ w \end{bmatrix} \Rightarrow (x/w, y/w)$$

Figure 28

# Block extraction

- At this step, we consider that the tray is well extracted. If the previous step failed, we can select manually the tray.
- The more crucial step consists in determining a good mask for the blocks.
- Once the mask is created, we find the contours on the mask. If the mask is correct, each contour will correspond to a block.

# First mask.

Many strategies for this mask but for this presentation we only consider the mask InRange based on HSV color space.

This mask is simply 3 low thresholds and 3 high thresholds for the 3 components of the HSV space.

The more efficient way consists in determining by hand these thresholds for a set of photos, because the conditions change for each sets.

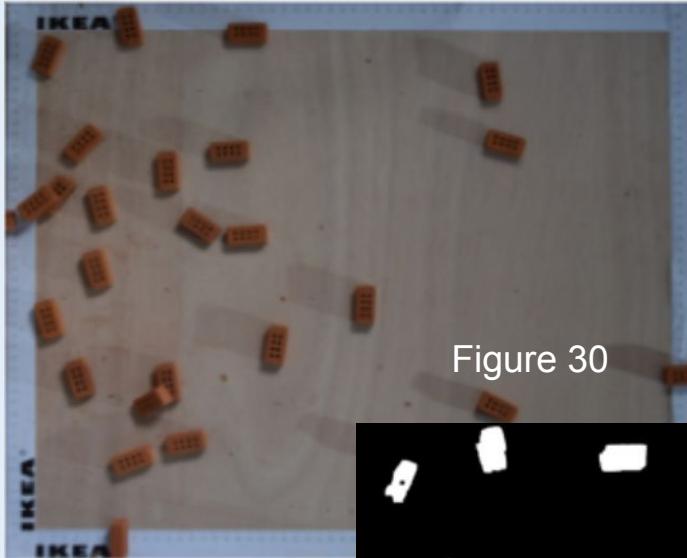


Figure 30

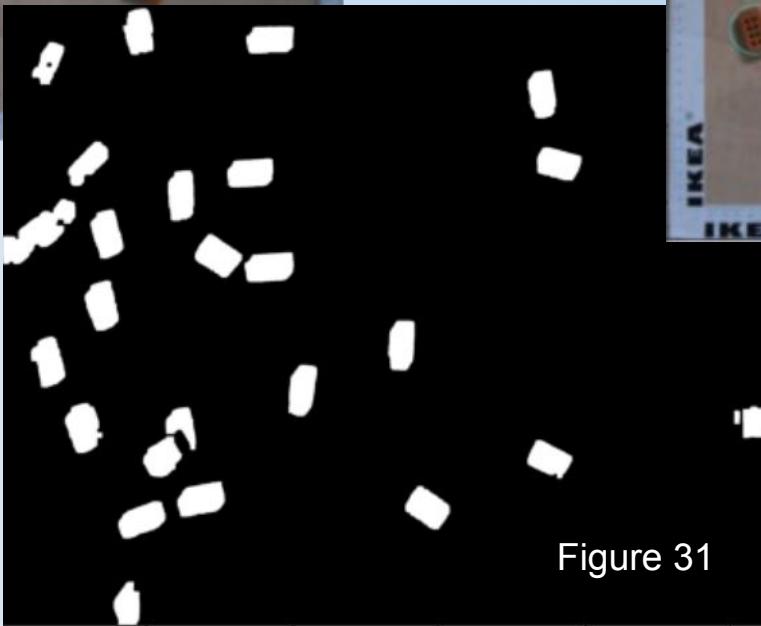


Figure 31

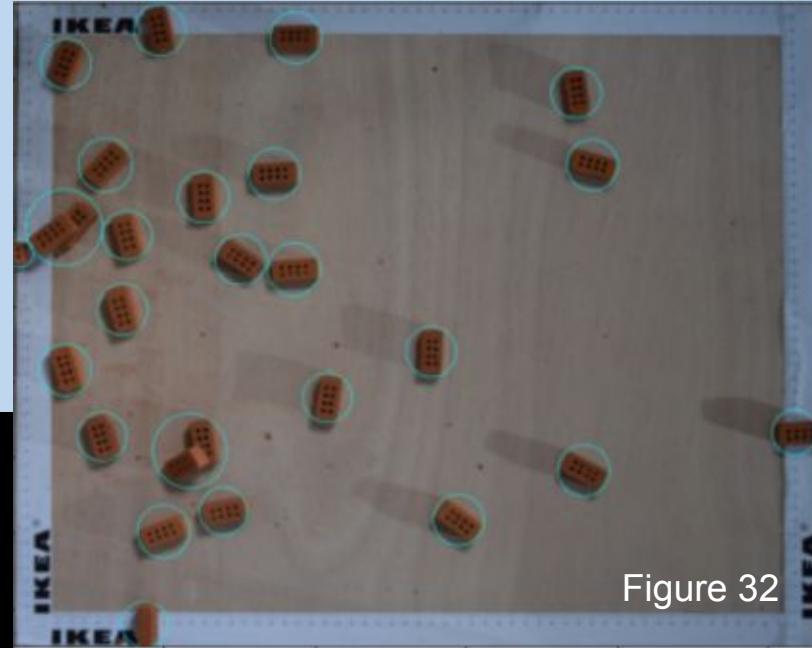


Figure 32

## Second mask

- The first mask is not sufficient because a circle may correspond to many blocks. So a new mask is necessary.
- The first mask is applied to the original image, then Canny edges detector allows us to find border between each block. We add these border to the previous mask.
- To increase the quality of the new mask, we can detect the little dark points of the blocks (thanks to Hough circle algorithm) and remove them from the new mask.

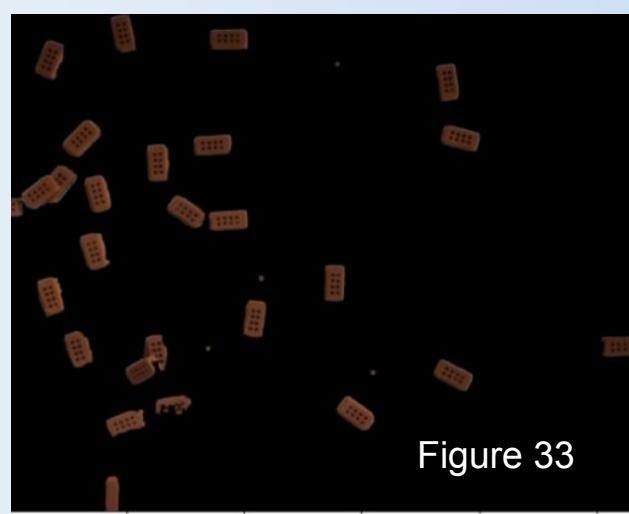


Figure 33

Canny

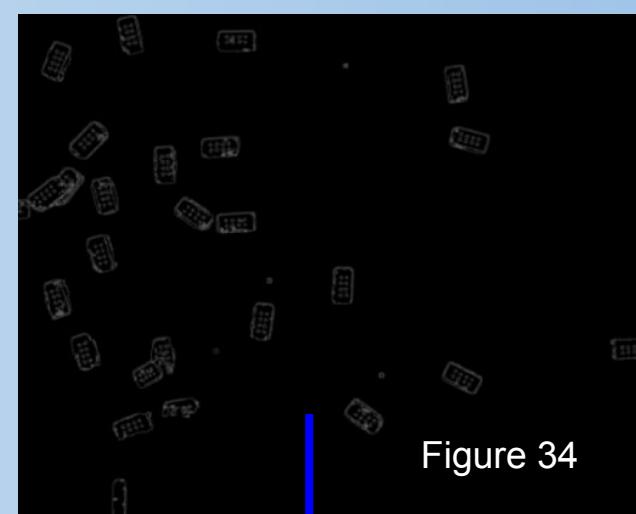


Figure 34

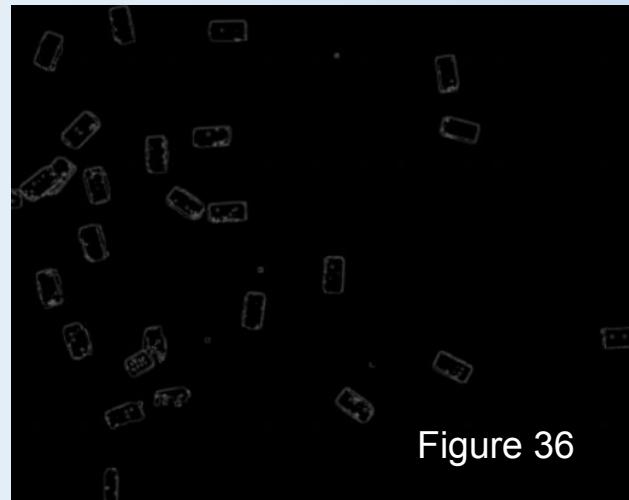


Figure 36

Circles deletion

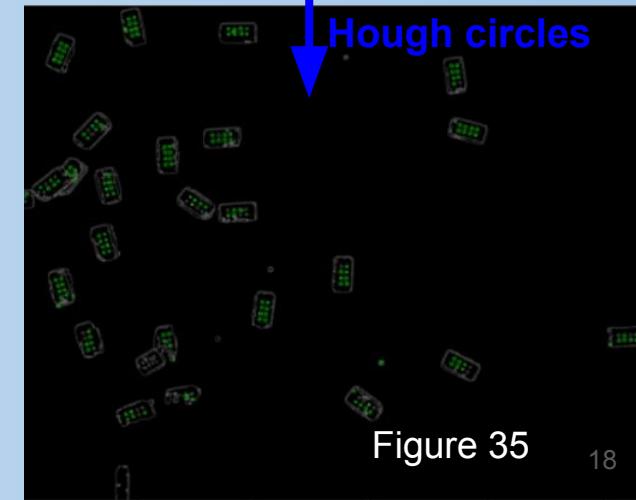


Figure 35

Hough circles

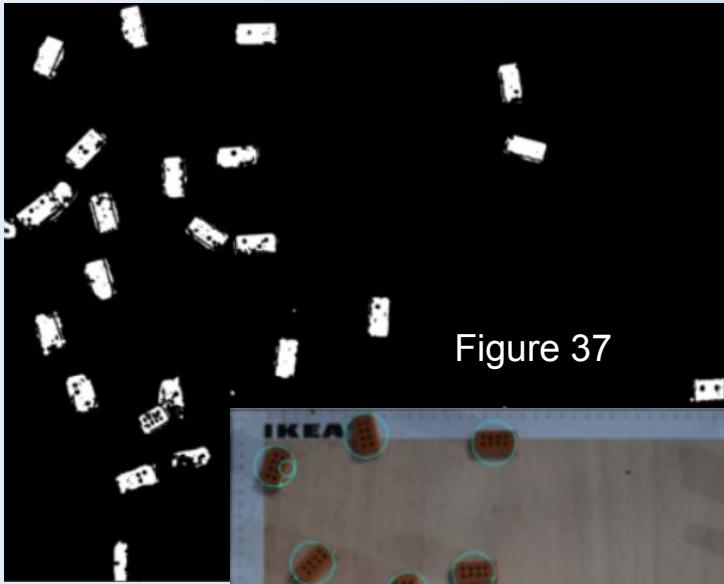


Figure 37

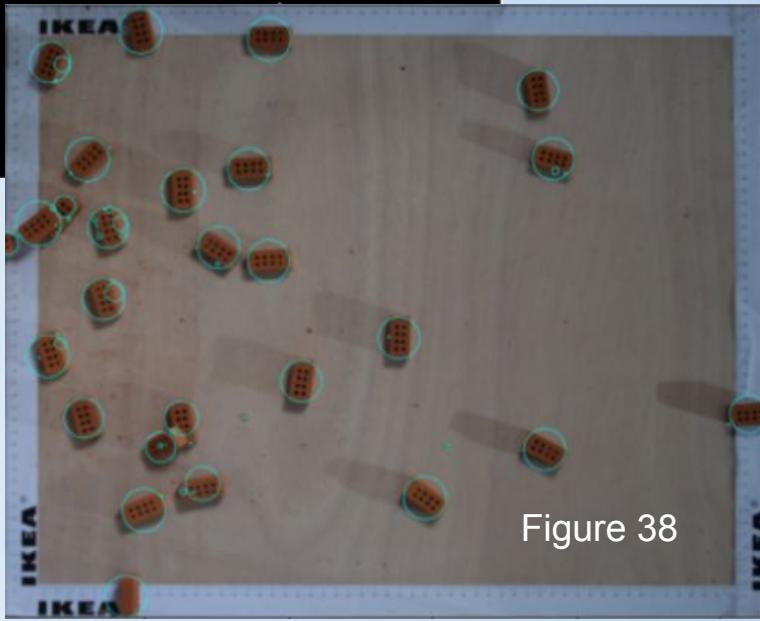


Figure 38

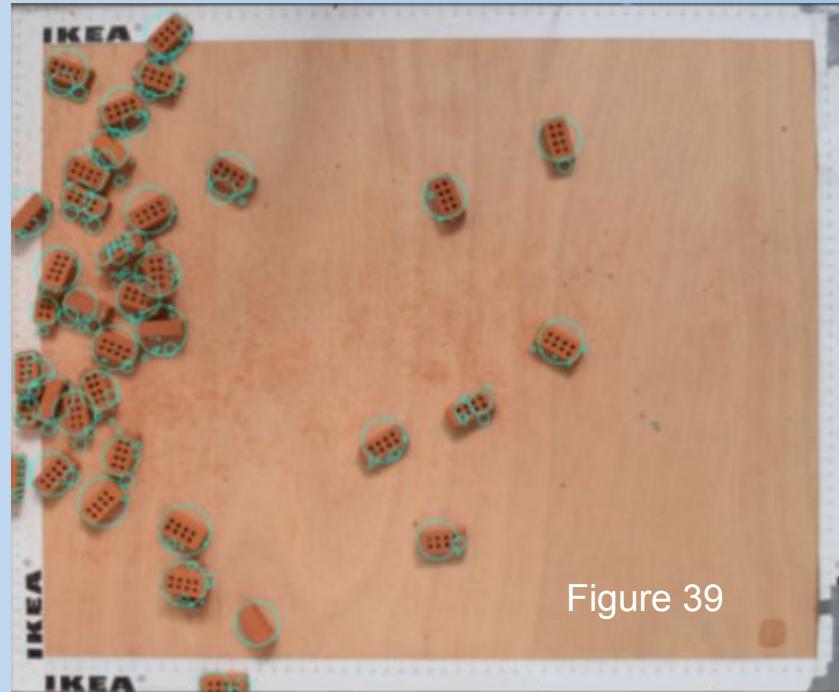


Figure 39

# Circles filtering

- The new filter make appear too many contours and therefore too many circles. However it appears that a clever filtering of them can lead to good results.
- Filtering :
  - Deletion of very little circles.
  - Deletion of circles inside a bigger circle.
  - Reunion of very near medium circles into a bigger circle.
  - If a circle exceed a limit radius, we consider than it contains 2 blocks.  
Thus we delete it and we take the 2 bigger circles which were inside it.

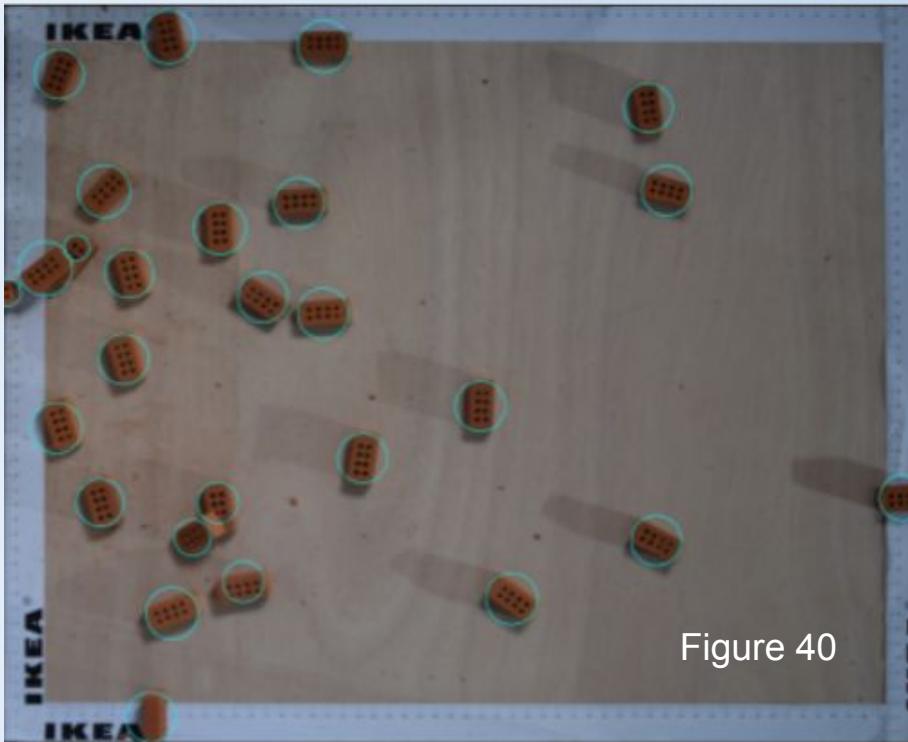


Figure 40

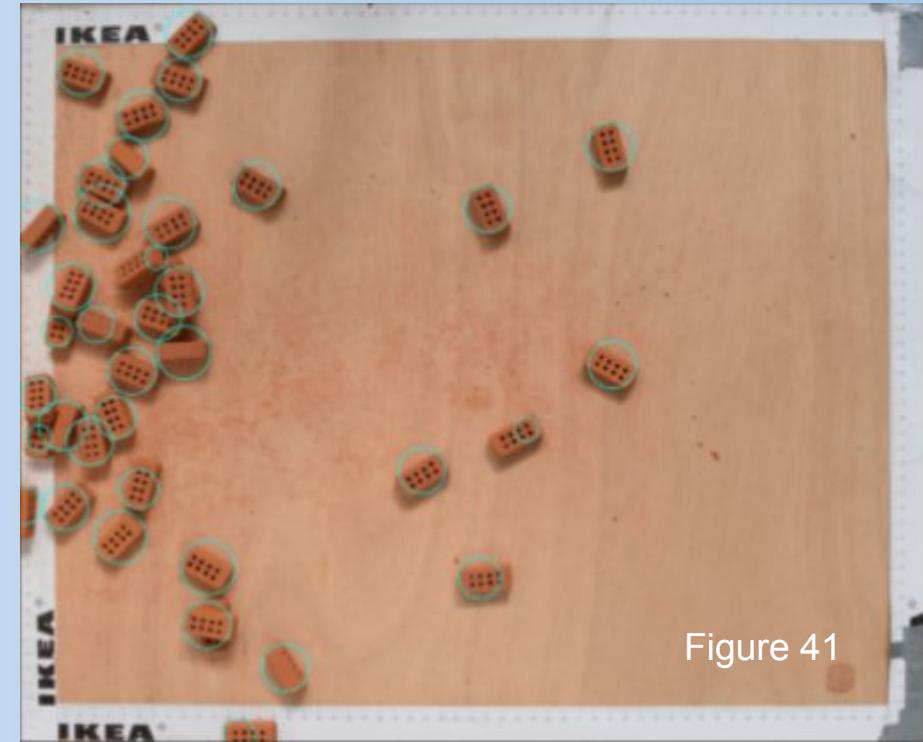


Figure 41

# Limitations

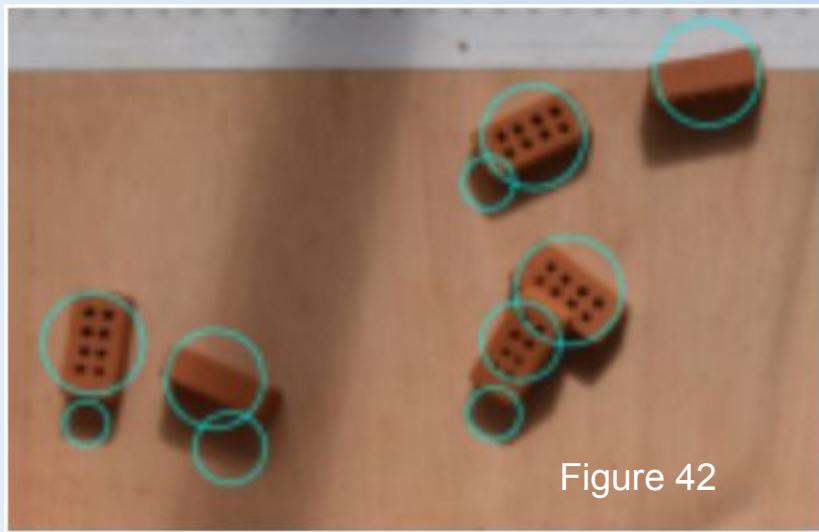


Figure 42

Shadows

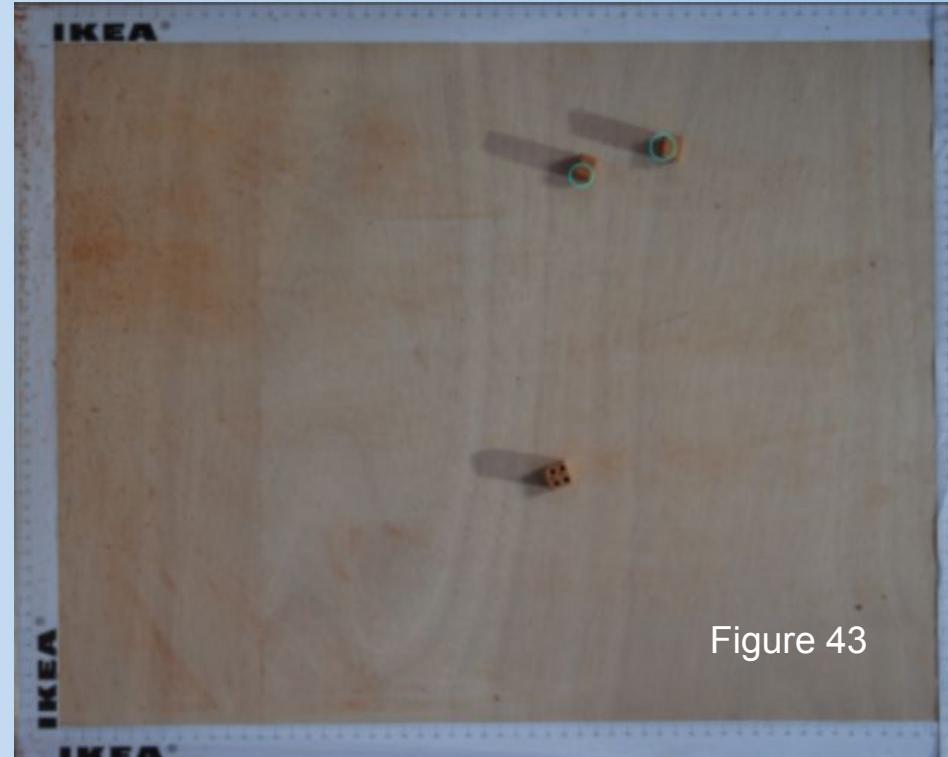


Figure 43

Low brightness

# Application

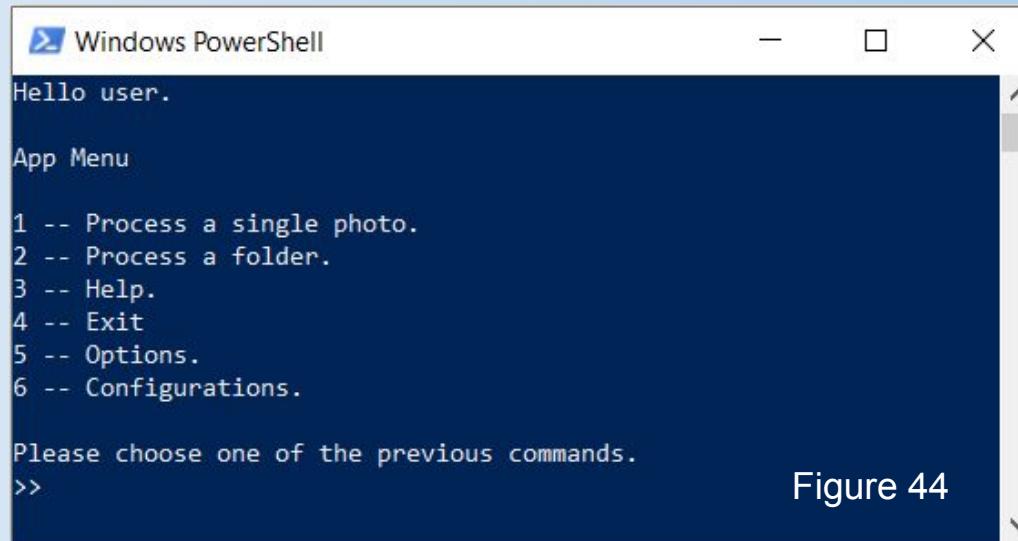


Figure 44

Menu of the app.

# Options

```
Windows PowerShell
Options

1 -- Change mask technic.
2 -- Path.
3 -- Manual/automatic threshold.
4 -- Return to the menu
More options will soon be available.

Please choose one of the previous commands.
>>
```

Figure 45

```
Windows PowerShell
Mask options.

1 -- Color mask.
2 -- Subtraction mask.
3 -- Adaptative mask.
4 -- HSV mask.

Please choose one of the previous commands.
>>
```

```
Windows PowerShell
Path

1 -- Automatic.
2 -- Manual

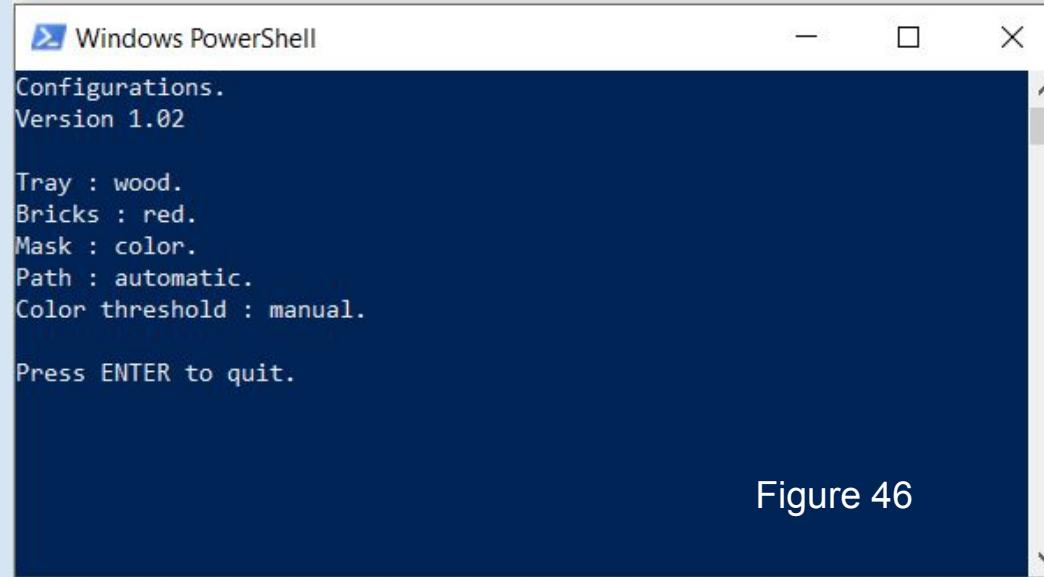
Please choose one of the previous commands.
>>
```

```
Windows PowerShell
Color threshold

1 -- Automatic.
2 -- Manual.

Please choose one of the previous commands.
>>
```

# Configurations



A screenshot of a Windows PowerShell window titled "Windows PowerShell". The title bar includes standard window controls: a minus sign for minimize, a square for maximize/minimize, and an X for close. The main content area displays the following text:

```
Configurations.  
Version 1.02  
  
Tray : wood.  
Bricks : red.  
Mask : color.  
Path : automatic.  
Color threshold : manual.  
  
Press ENTER to quit.
```

The window has a dark blue background and white text. A vertical scroll bar is visible on the right side of the window.

Figure 46

You can check the current options in configurations.

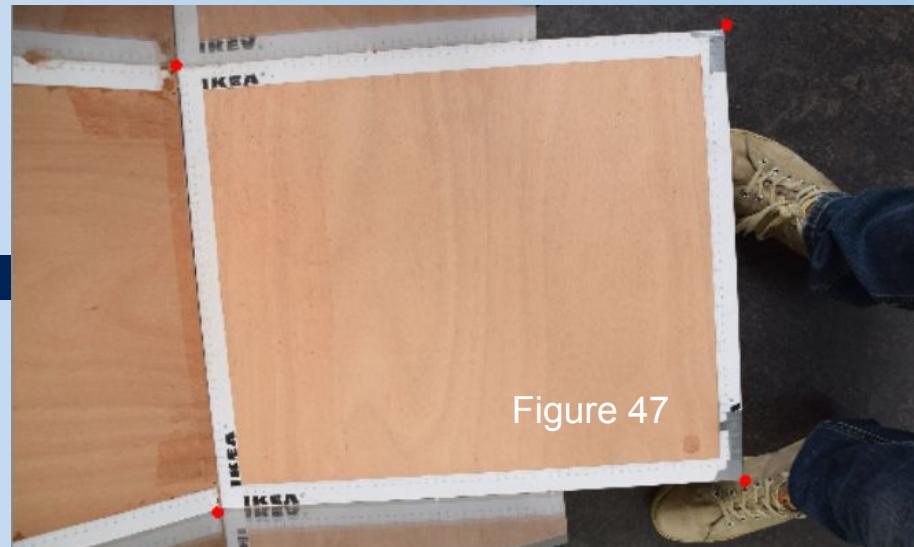
# Process a folder

```
Sélection Windows PowerShell  
Which folder ? (Please write the path to the folder you want to process.)  
>> ../../stage_simon_photos/14 briques/grosses briques/bois/Photos
```

The app tries to detect the 4 angles and ask the user to confirm or infirm.

```
Confirm ? y/n
```

If the user refuse, he can click himself on the 4 angles.



Corners successfully saved.

You have to choose manually the color to detect.

Please click at least 10 times on bricks.

In the case of manual threshold,  
the user must click on the bricks  
so that the app know the color of  
the bricks.

Then the algorithm will both save  
photographies with detected blocks  
and save the associated csv file.

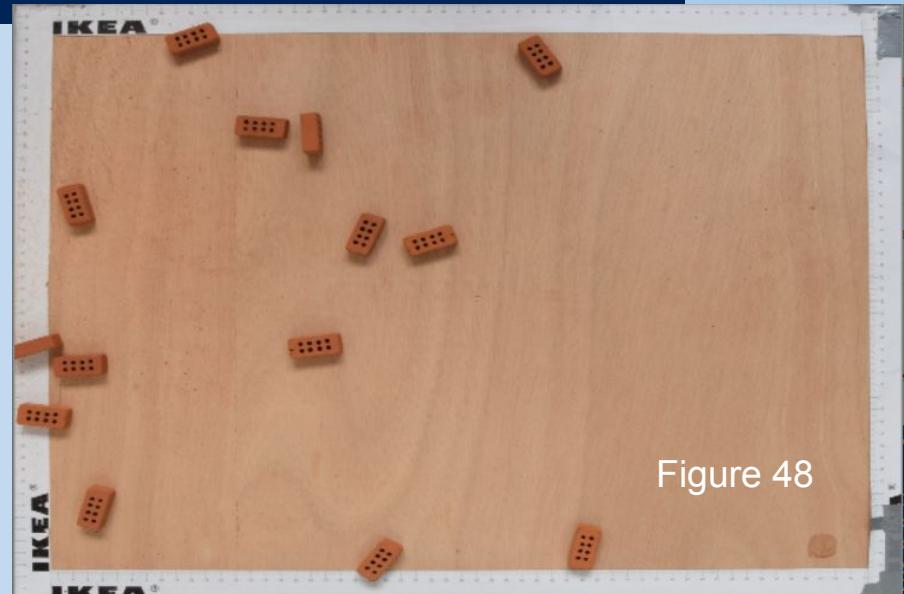
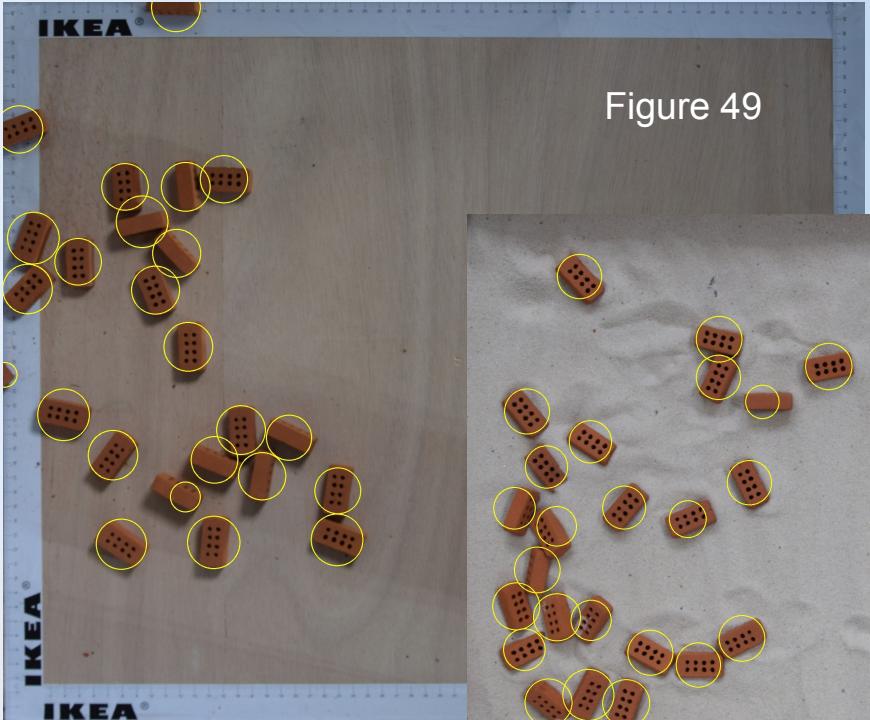


Figure 48

# Some results



# Results and Recommendation for the shooting

Our algorithm was evaluated on 50 pictures. On each pictures 14 big blocks were thrown on a wooden tray.



Figure 52

# Data

0	1	2	3	4	5	6	7	8	9	...	21	22	2	
0	1	14	0	35.197523	16.519179	19.998295	11.941713	4.513883	23.599115	27.769780	...	17.336349	19.910980	20.53501
1	2	14	0	31.298408	27.890697	41.458567	25.225096	21.888791	22.081376	31.207467	...	31.524062	33.083644	43.24920
2	3	14	1	36.467089	28.325117	19.641443	17.431282	14.230665	11.643330	10.852407	...	10.345129	20.140925	27.18834

Columns :

- 0 represents an index
- 1 represents the number of block thrown
- 2 represents the number of block outside the tray
- 3 - 16 X-coordinates
- 17-30 Y-coordinates

Figure 31

# Number of block outside the tray

- 66% of accuracy
- 33 pictures

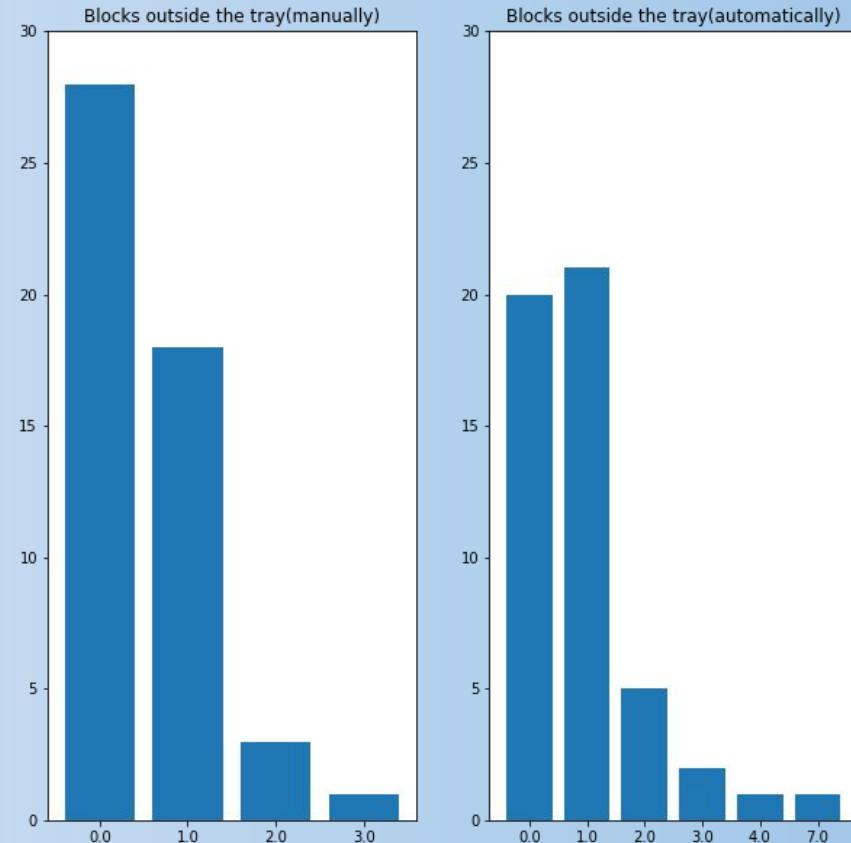
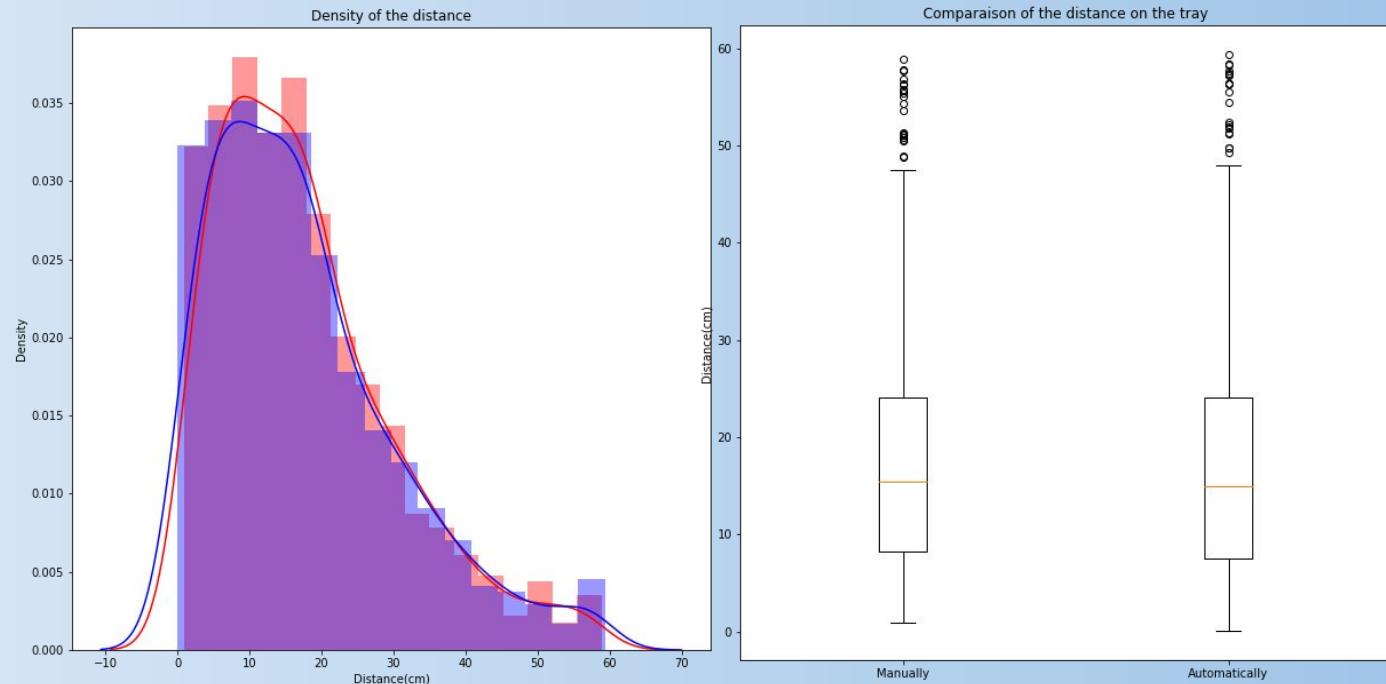


Figure 54

# Influence on the distance of the automatic measurement

Data : Concatenation  
of the concatenation of  
X-coordinates for each  
image without the 0.



The two dataset  
follows the same  
distribution.

Figure 55

Figure 56

# Analysis of the error

During the error analysis we will only look at the images where the right number of bricks have been found.

Data:

- Mean : 0.33
- Variance : 0.58

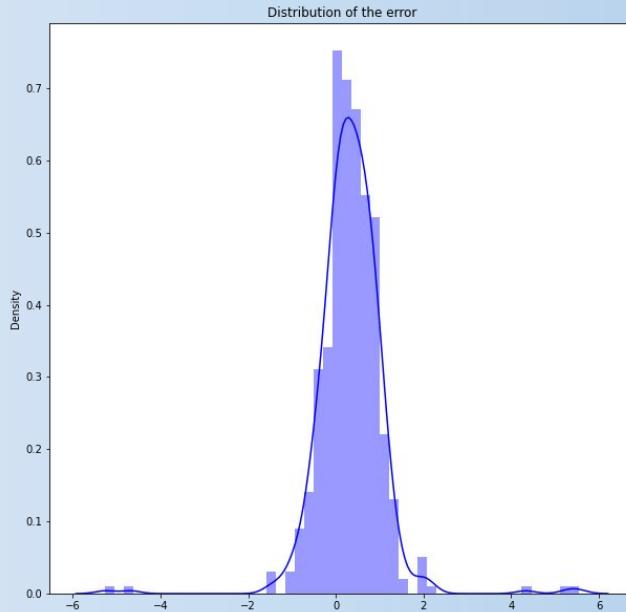


Figure 57

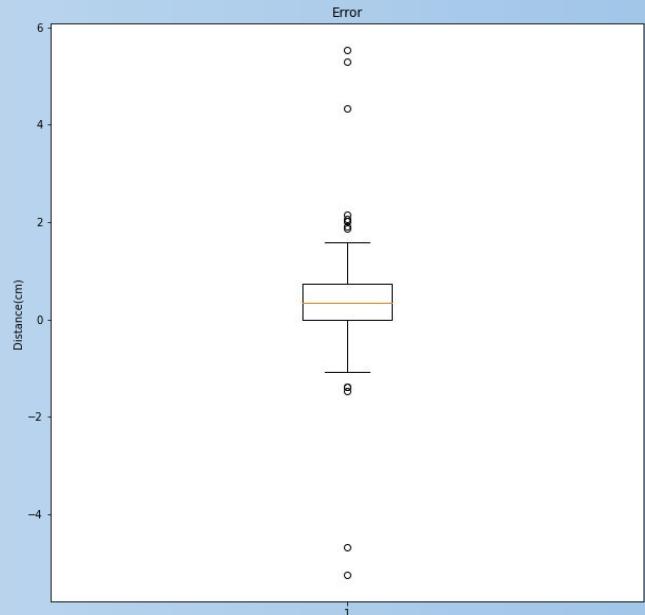


Figure 58

# Analysis of the error for each picture

Mean absolute error = 0.73 cm

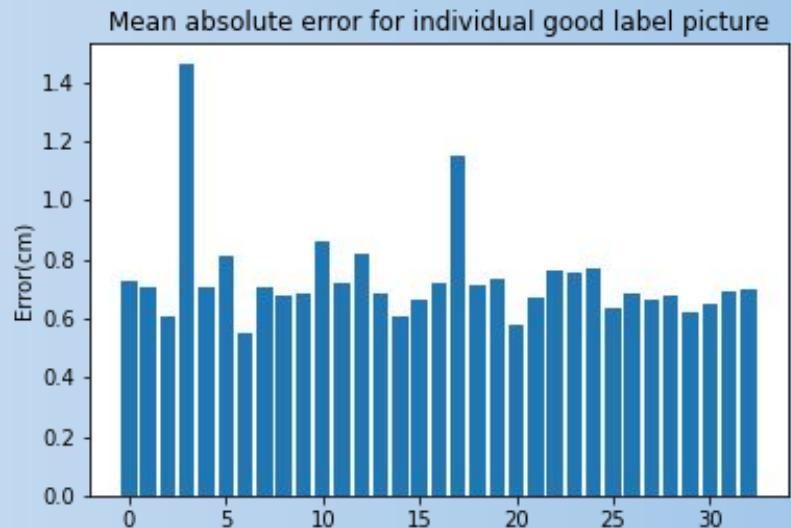


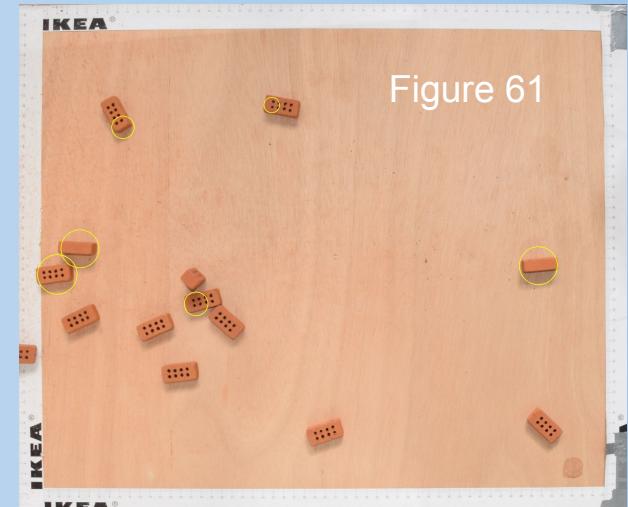
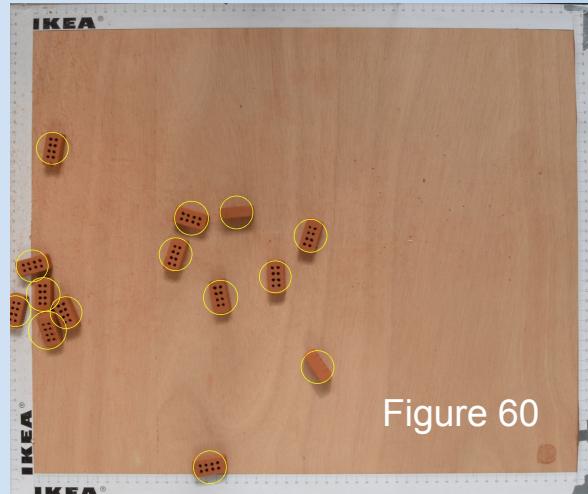
Figure 59

# Recommendation for shooting

We have given a lot of comments and advice in the report on how to take the pictures to improve the results. In this last part, we will only give in relation to the shooting done on the previous dataset.

# Recommendation : Environment

The brightness should  
be the same for all  
images in the same  
folder.



No objects other than bricks must appear on the tray.

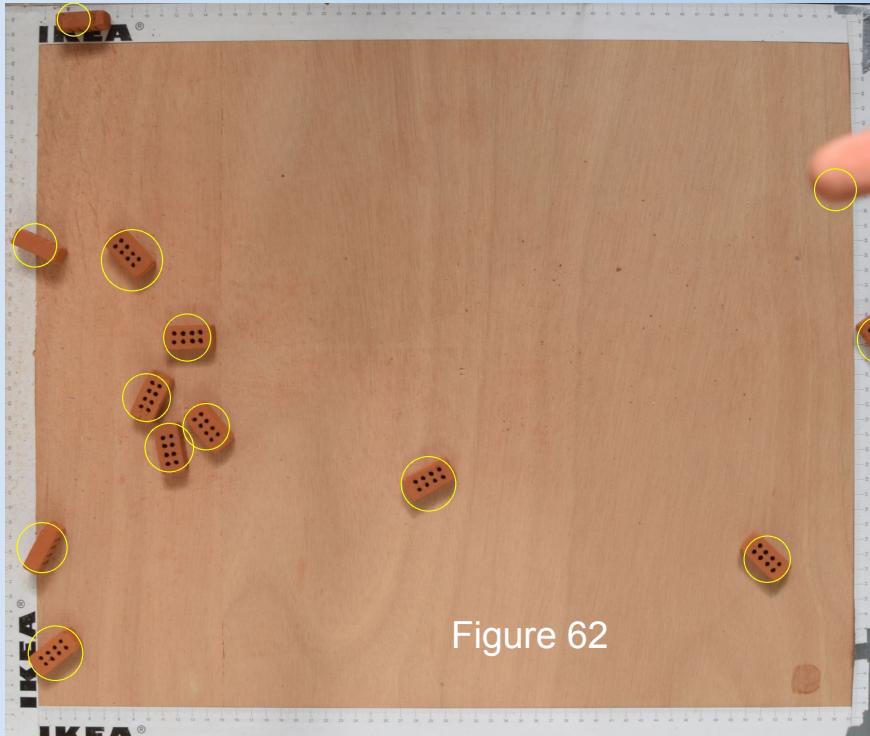


Figure 62

Thank you for your time and your attention