

A simple detector for crowd counting using OpenCV and Python

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1 Introduction

The assignment consists in **counting** the number of people that appear on the given images. There was not an established way to approach this problem. In this case I have taken the decision of using the computer vision algorithms seen in class and not rely on deep learning or any kind of artificial intelligence algorithm. This is because not by the amount of data needed for training, due to there are really good datasets out there, like MS COCO, but because of the effort of building a good performing detector architecture. Use an already set framework as YOLO or Mask-RCNN would be the way to accomplish the task, but I would be thinking about it as "cheating". Therefore, this way, using the content seen in class, I will be more cautious about my decisions and it will be much easier to justify them.

In the following document we will be focusing on the process taken and the algorithms used, their implementation and their performance.

2 procedure

The program takes the following steps :

1. Import images as black and white.
2. Apply Adaptive Histogram equalization to the images.
3. Subtract the background to the images using the image with the background.
4. Apply an erosion algorithm to obtain small highlighted areas.
5. Apply a thresholding algorithm to binarize the image.
6. Apply a dilation operation into the binarized images to expand the whites.

7. Use a contour algorithm to extract the different regions containing persons.
8. Obtain the bounding boxes from these regions.
9. Count them and compare the number of detections to the real quantity.

3 algorithms

In this section we will discuss the algorithms selected and the output we obtain from them in the application of our problem.

3.1 Background Removal

Background removal is a technique that allows to remove the background from the image, this way, the output will be only the foreground highlighted in the form of a binarized image as it can be illustrated on the figure 1.

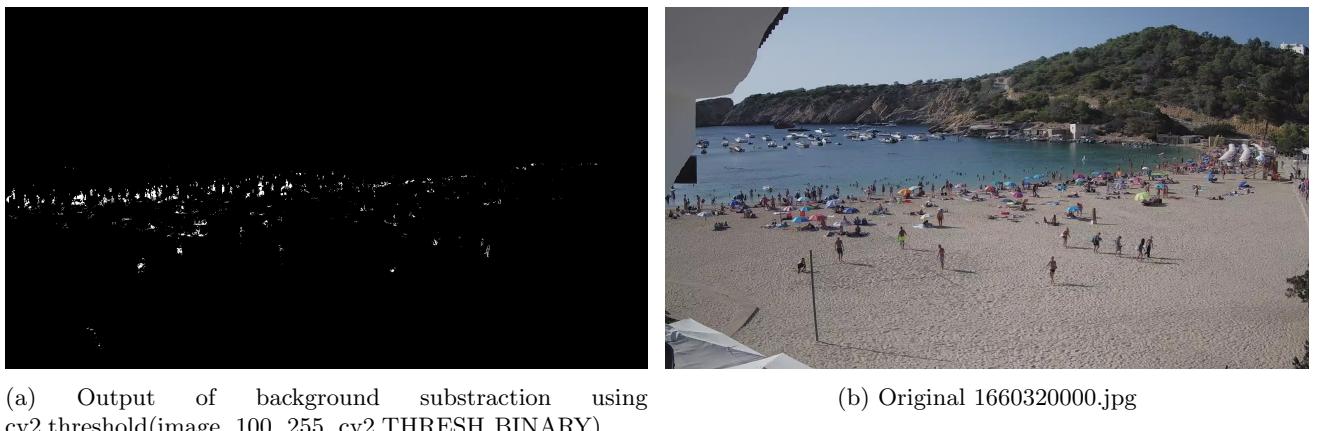


Figure 1: Reference images

In the following sections will be explained how the preprocessing of this technique was prepared and for what they are used for. They are ordered by

CLAHE

In the case of our problem, first we have imported all the images in black and white integer values using `cv2.imread(img,0)`. Then for a better extraction, it has been applied an **algorithm of CLAHE** seen previously on the course to try to uniformize the illumination of the images. With a more uniform images, we will be able to distinguish more precisely our foreground and artifact like shadows casted by objects on the images may have less impact on the **binarization process**.

Background Image

For the selection of our background image, in a starting point the **average image** was computed as the main way for accomplish foreground extraction, but the resulting image generally was really noisy and with ghosting effect very present on it, which decreased the performance of the algorithm as it can bee seen on the Figure 3. Even with a gaussian blur applyed to try to counter the ghosty figures, they are still really present on the image. The image averaging is a good idea generally for noise reduction and in this case to try to obtain a neutral image were the background is predominant, but there is not enought images to obtain a more sharp image to really have a good background. Therefore, we **end up** using the image `1660284000.jpg` with a gaussian blur applied as our background, this way we obtain better results on the **binarization process**.

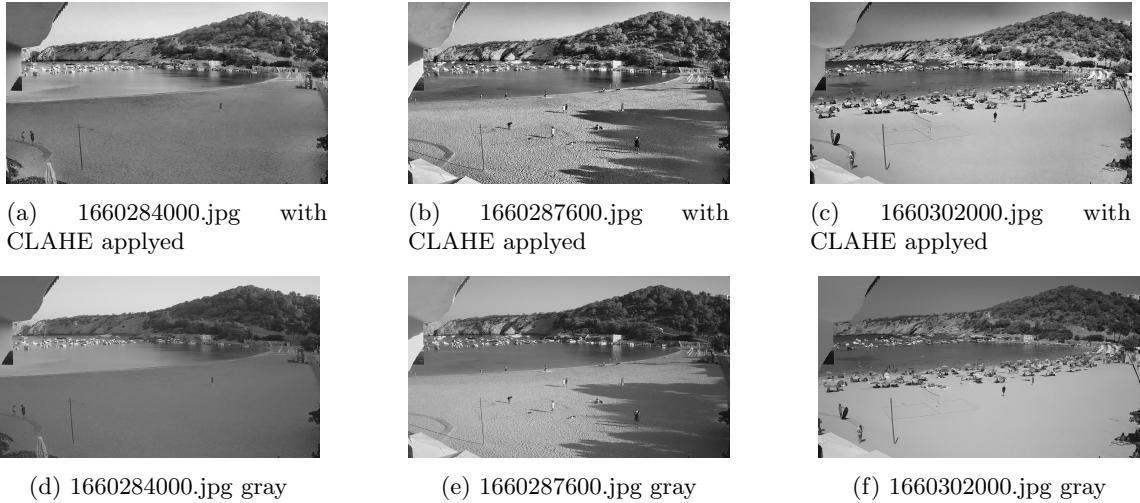


Figure 2: Images from Gelabert folder where the first row presents a more uniform ilumination across frames ompared with the second row images.

Gaussian Blur

With an image already selected for the image background, it is applyed a gaussian filtering with a kernel of size (7,7), this way we can obtain a more standard image, and get a more smuthered image that will allow the algorithm to contrast slightly better our foreground.

The ending size used for gaussian blur was obtained by trial and error experimentation, and could be changed to get different ending behaveours.

Substraction

Once we have selected our background and applyed the preprocessing we should substract every image with our background, this way, the resulting image is the foreground (in this case the persons and some residuals) highlighted as we can see in the figure 5.

THRESH_OTSU funciona bien salvo en los casos donde empieza a haber compras en la imagen, a la hora de aplicar el thresholding obtendremos un n mero muy grande de falsos positivos en las peque as monta as que hace la arena, las dimensionas que tienen son similares a las de una cabeza humana por lo que si las eliminasemos estariamos eliminando buenas detecciones.

3.1.1 Image Averaging

3.1.2 Image Substraction

4 Gabor Filter

A band pass filter generated by a function of various parameters.

$$filter(x, y; \sigma, \theta, \lambda, \gamma, \phi) = \exp\left[-\frac{x^2 - \gamma^2 \cdot y^2}{2\sigma^2}\right] \cdot \exp[i(2\pi\frac{x}{\lambda} + \phi)] \quad (1)$$

The parameters of ksize allows to select the size of our kernel filter, in case we are using a really big shape we will overlook details if the shapes are small. The same reasoning can be applyed with a small filter, may overlook shapes too big for it. Therefore, must be tested with different sizes to reach an idoneal spot, if your features are tiny or bigger, you have to take that in count.

If we are looking for **horizontal-like** features, applying an horizontal filter will allow us to maintain those characteristics and block the vertical ones and viceversa in the other cases.



Figure 3: Average image with a gaussian blur applied with a kernel of (11,11)



Figure 4: Gaussian Blur Applied with a kernel of size (7,7)

5 What form do the People have?

In general the shapes that a person can describe may suffer a lot of distortion depending of the angle where the frame was took, the person posture, etc. Not always will be a perfect pose to the researchers to easily identify if an object is a person or not. There is no key shape that we could use, but we can try use common sense. The images gathered, are related to people moving standing up, swimming or just taking sunbathing.



Figure 5: Image obtained by applying `cv2.subtract(background, image)`



Figure 6: Binarized image using OTSU and the empty beach image