Visual Perception

Practical Lab: Corner Detection

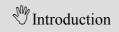
Related to Lecture 3 Workload: 4 hours in the lab Programming platform: Matlab

1. Objective

Experience on how to detect corners in an image.

Understand the importance of the different steps involved in corner detection using the Harris algorithm.

2. Corner detection



In this practical session we are going to implement the well-known **Harris** corner detector.

This method is based in the use of the image gradient. The procedure is as follows:

Compute the derivatives of the intensity image in the x and y directions for every pixel.

In this way we obtain $I_x=rac{\partial I}{\partial x}$ $I_y=rac{\partial I}{\partial y}$ for every point of the image.

Then, we compute the products:

$$I_x^2$$
, I_y^2 and I_xI_y

Next, the squared image derivatives should be smoothed by convolving them with a Gaussian filter g.,

$$\left\langle I_{x}^{2}\right\rangle =g\otimes I_{x}^{2}\tag{1}$$

where \otimes is the convolution operator, and the autocorrelation matrix ${\bf M}$ can be defined for every pixel:

$$\mathbf{M} = \begin{pmatrix} \sum_{W} \left\langle I_{x}^{2} \right\rangle & \sum_{W} \left\langle I_{x} I_{y} \right\rangle \\ \sum_{W} \left\langle I_{x} I_{y} \right\rangle & \sum_{W} \left\langle I_{y}^{2} \right\rangle \end{pmatrix}$$
(2)

where W is a 3×3 neighborhood around the point.

Remember that M can be used to derive a measure of "cornerness" for every pixel. Depending of the rank of M, the pixel belongs to an homogeneous region (rank M=0), an edge (significant gradient in 1 direction, that is, rank M=1), or a corner (significant gradients in both directions, thus, rank M=2).

Harris and Stephens proposed the use of the determinant and the trace of ${\bf M}$ to detect corners:

$$R = \det(M) - k \cdot tr(M)^2 \tag{3}$$

where: k is a constant (with a value of 0.04)

 $det(\cdot)$ is the determinant and

tr is the trace (sum of the diagonal elements).



Go to the course web page and download the files <code>corner_detector.m</code> and the image <code>chessboard00.png</code>. The matlab script is just the starting point from which you should carry out this practical session.

```
im = imread('chessboard00.png');
im oriq = im;
if size(im,3)>1 im=rgb2gray(im); end
% Derivative masks
dx = [-1 \ 0 \ 1;
       -1 0 1;
       -1 0 1];
dy = dx';
% Image derivatives
Ix = conv2(double(im), dx, 'same');
Iy = conv2(double(im), dy, 'same');
sigma=2;
\mbox{\ensuremath{\$}} Generate Gaussian filter of size 9\mbox{\ensuremath{\$}} and std. dev. sigma.
g = fspecial('gaussian',9, sigma);
\mbox{\%} Smoothed squared image derivatives
Ix2 = conv2(Ix.^2, g, 'same');
Iy2 = conv2(Iy.^2, g, 'same');
Ixy = conv2(Ix.*Iy, g, 'same');
```

Part 1

Modify the code above to compute a matrix $\rm E$ which contains for every point the value of the smaller eigenvalue of $\rm M.$ Once computed, display this matrix by means of:

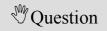
```
imshow(mat2gray(E))
```

Part 2



Modify the code above to compute a matrix R which contains for every point the result of Equation (3). What is the difference with respect to E? Use functions tic and toc to measure the time required for computing E and R.

Part 3



Select for E and R the 81 most salient points. Is this the result you expected?

Hint for part 3: Build a structure named features with 2 fields to store the x coordinate (p x) and the y coordinate (p y) of the 81 points with the highest

the cornerness value given by E or R. Then, display the corners by means of:

```
figure; imshow(im_orig); hold on;

for i=1:size(features,2),
        plot(features(i).p_x, features(i).p_y, 'r+');
end
```

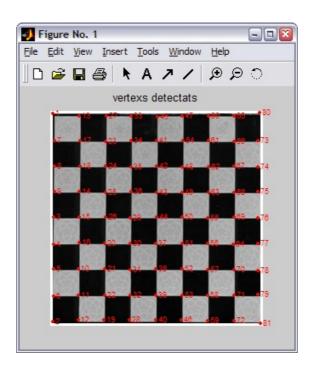
Test your code with images chessboard01.png, chessboard02.png, etc.

Part 4

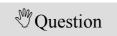


Build a function to carry out non-maximal suppression for E and R. Again, the 81 most salient points using a non-maximal suppression of 11×11 pixels. Is this result better than that of Part 3?

Test your code with images chessboard01.png, chessboard02.png, etc.



Part 5 (this is an optional extrapart)



Find the location of the corners with subpixel accuray. What function would you fit on top of the cornerness values?