

# Image derivatives, edges, **corners**, . . .

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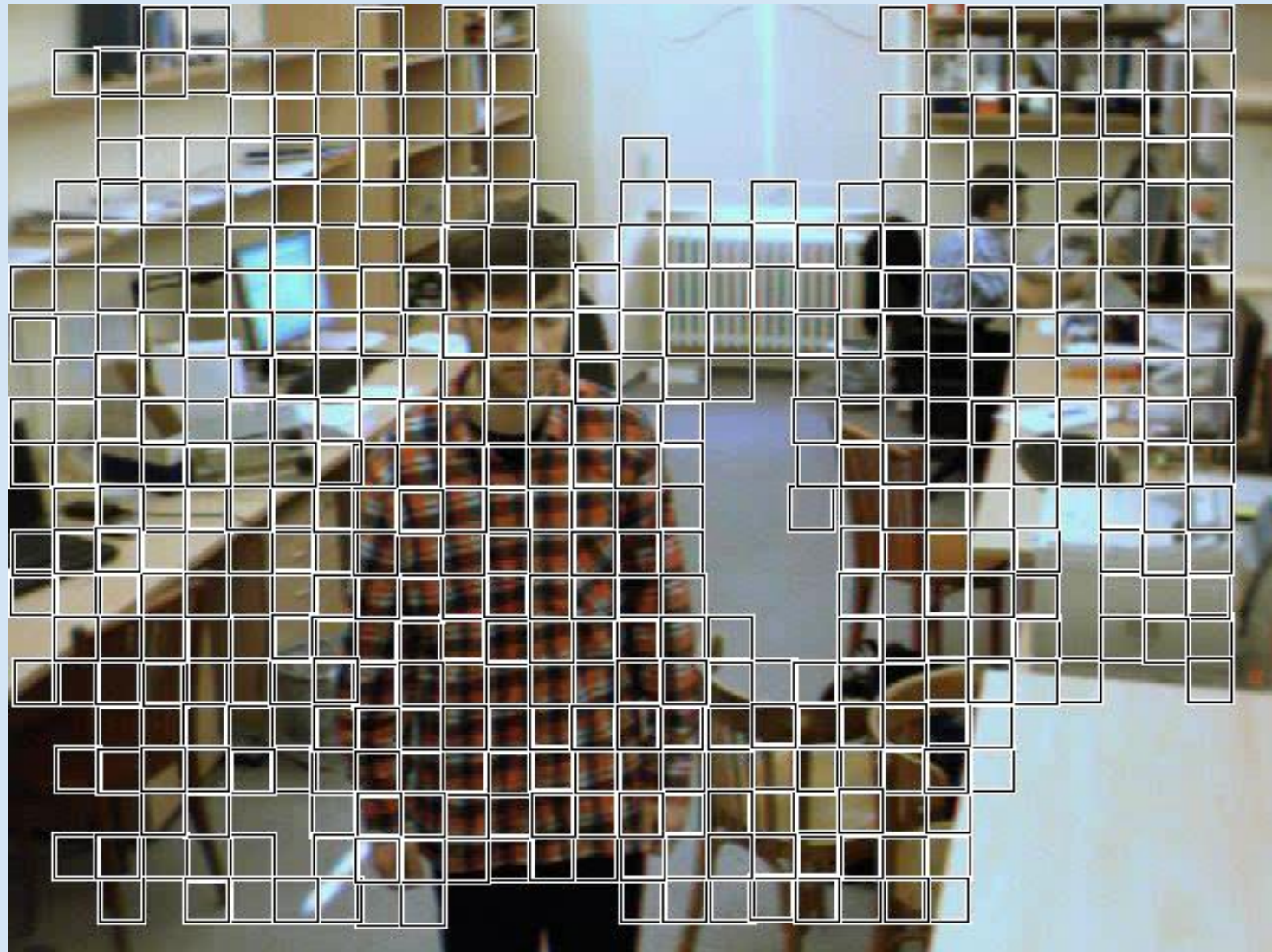
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**Last update:** October 26, 2009

## Talk Outline

- ◆ why corners?
- ◆ combined edge and corner detector
- ◆ Demo: step by step in Matlab

# Dense tracking



Video: Dense tracking. Note that the coverage is not complete

# What are good features (windows) to track?

How to select good templates  $T(\mathbf{x})$  for image registration, object tracking.

$$\Delta \mathbf{p} = \mathbf{H}^{-1} \sum_{\mathbf{x}} \left[ \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \right]^\top [T(\mathbf{x}) - I(\mathbf{W}(\mathbf{x}; \mathbf{p}))]$$

where  $\mathbf{H}$  is the Hessian matrix

$$\mathbf{H} = \sum_{\mathbf{x}} \left[ \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \right]^\top \left[ \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \right]$$

The stability of the iteration is mainly influenced by the inverse of Hessian. We can study its eigenvalues. Consequently, the criterion of a good feature window is  $\min(\lambda_1, \lambda_2) > \lambda_{min}$  (texturedness).

# What are good features (windows) to track?

Consider translation  $\mathbf{W}(\mathbf{x}; \mathbf{p}) = \begin{bmatrix} x + p_1 \\ y + p_2 \end{bmatrix}$ . The Jacobian is then

$$\frac{\partial \mathbf{W}}{\partial \mathbf{p}} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned} \mathbf{H} &= \sum_{\mathbf{x}} \left[ \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \right]^\top \left[ \nabla I \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \right] \\ &= \sum_{\mathbf{x}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{\partial I}{\partial x} \\ \frac{\partial I}{\partial y} \end{bmatrix} \left[ \frac{\partial I}{\partial x}, \frac{\partial I}{\partial y} \right] \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= \sum_{\mathbf{x}} \begin{bmatrix} \left( \frac{\partial I}{\partial x} \right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left( \frac{\partial I}{\partial y} \right)^2 \end{bmatrix} \end{aligned}$$

The image windows with varying derivatives in both directions.

Homogeneous areas are clearly not suitable. Texture oriented mostly in one direction only would cause instability for this translation.

# What are the good points for translations?

The Hessian matrix

$$H = \sum_{\mathbf{x}} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix}$$

Should have large eigenvalues. Harris and Stephens were motivated differently, yet they ended with their famous<sup>1</sup> corner detector [1].

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<sup>1</sup>3635 times cited according to Google scholar on 2009-10-26.

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**Few questions arise:**

- ◆ why corners?
- ◆ what was the motivation for corners?
- ◆ what are corners, actually?

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# Image matching

- ◆ uniqueness
- ◆ repeatability

Derivation of the corner detector on the blackboard. Details can be found in [1]. Its implementation with demonstration of use in chapter 5 of [2].

# Image derivatives

In discrete space we approximate them by local differences:

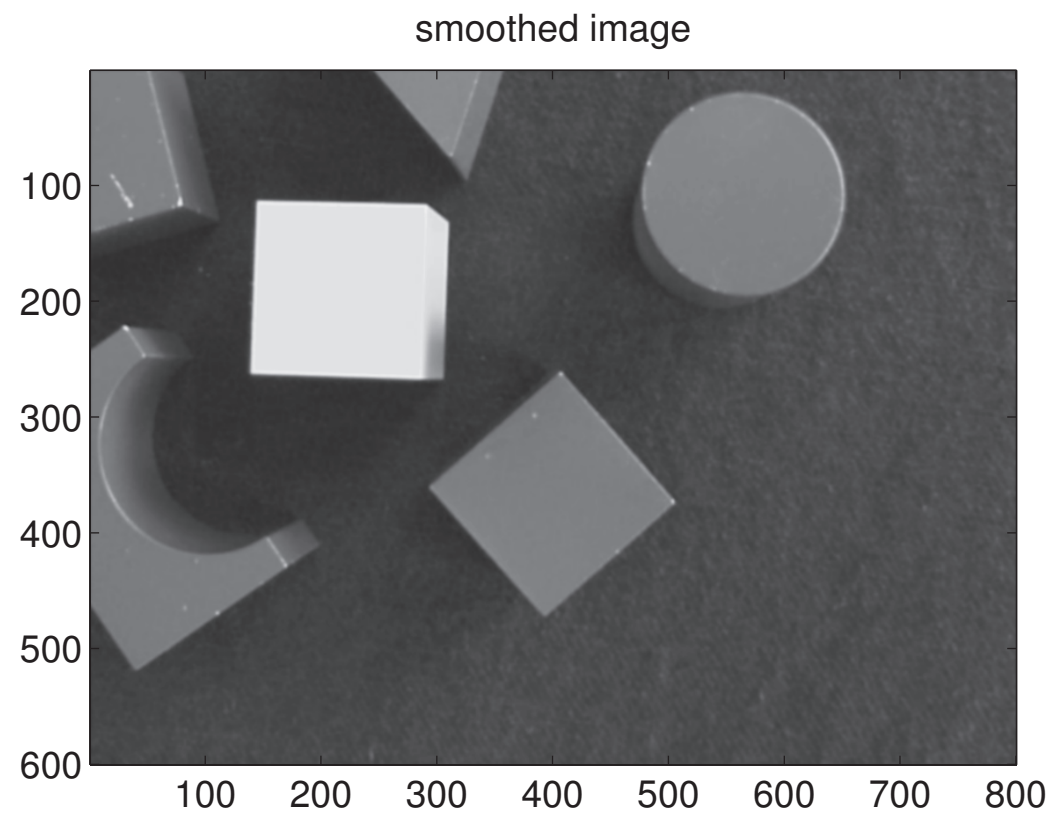
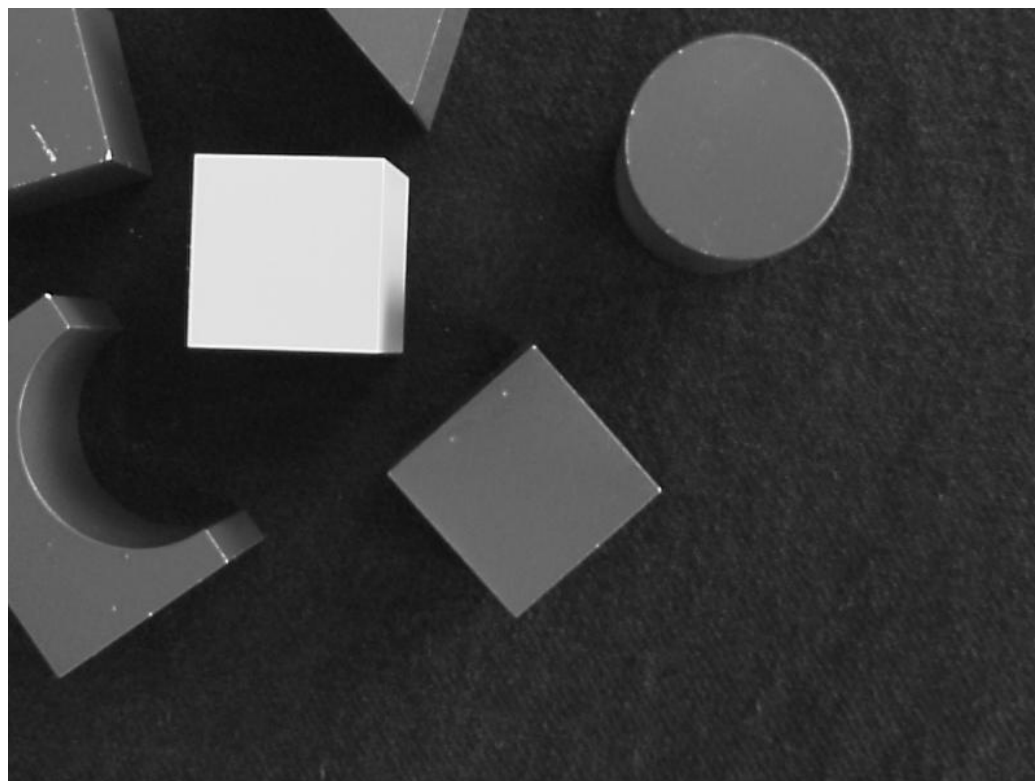
$$\frac{\partial I}{\partial x}(x, y) = I(x + 1, y) - I(x - 1, y)$$

$$\frac{\partial I}{\partial y}(x, y) = I(x, y + 1) - I(x, y - 1)$$

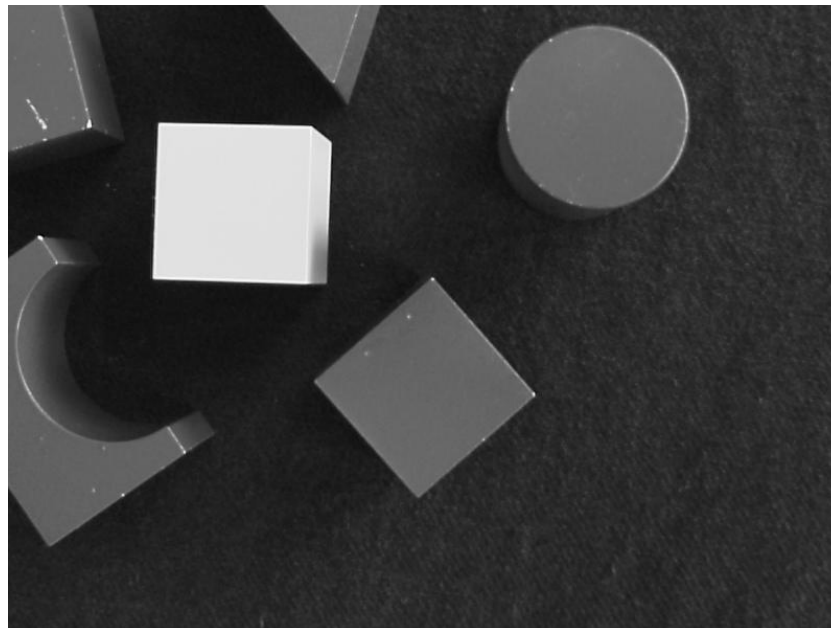
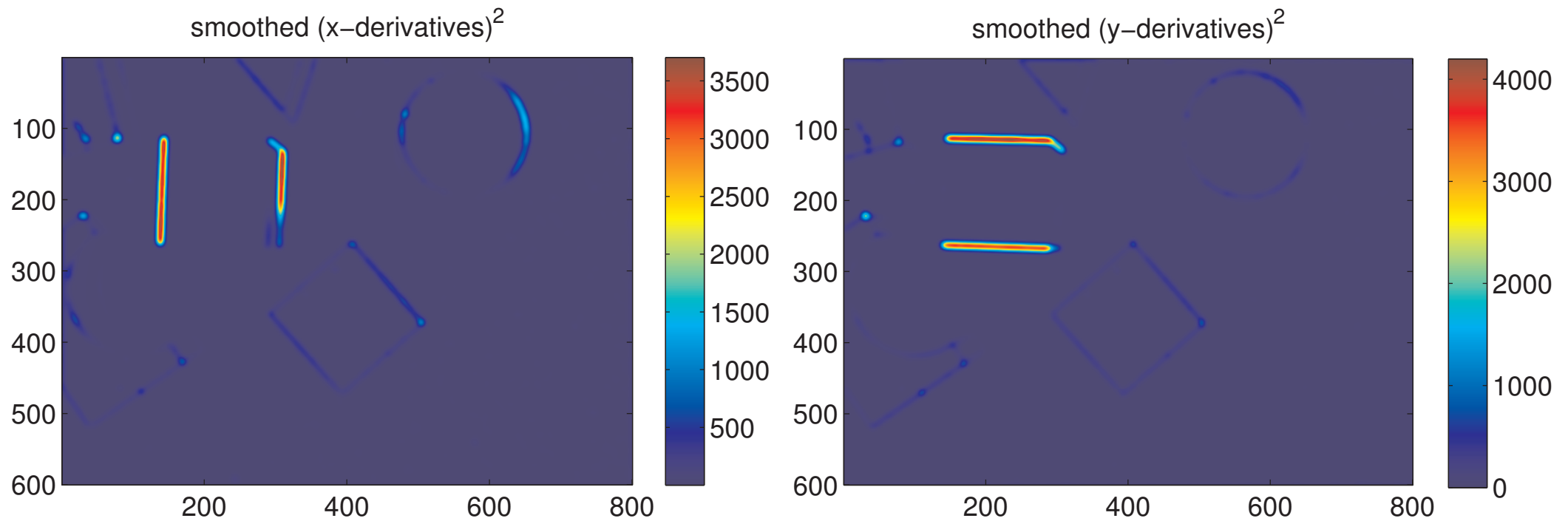
See the on-line demonstration in Matlab



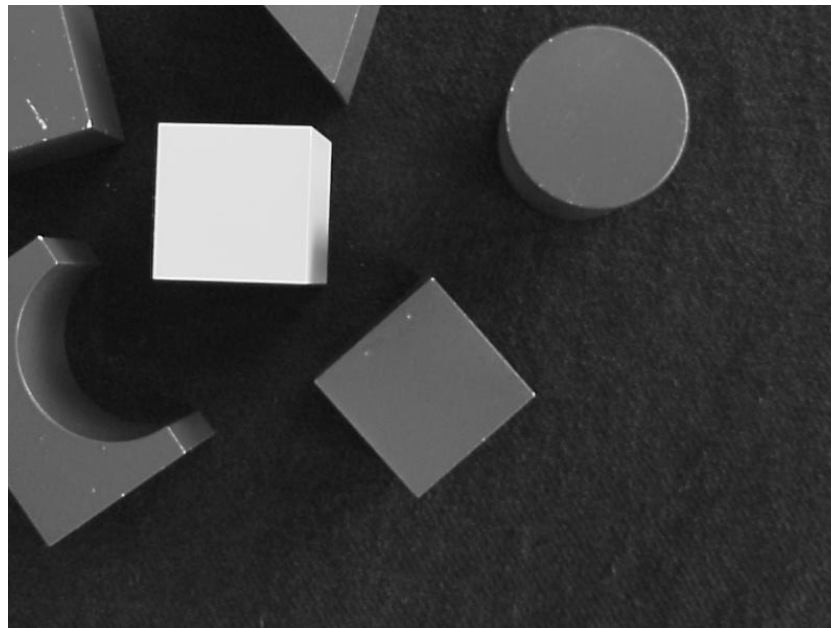
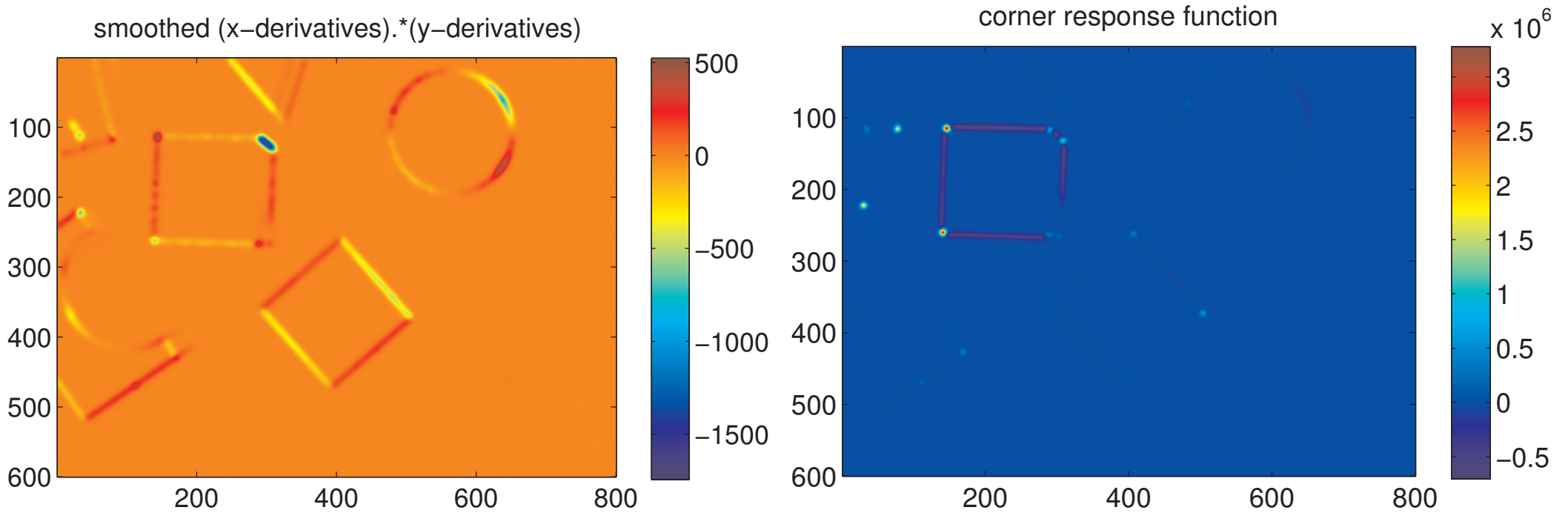
# Corner detection in pictures, images



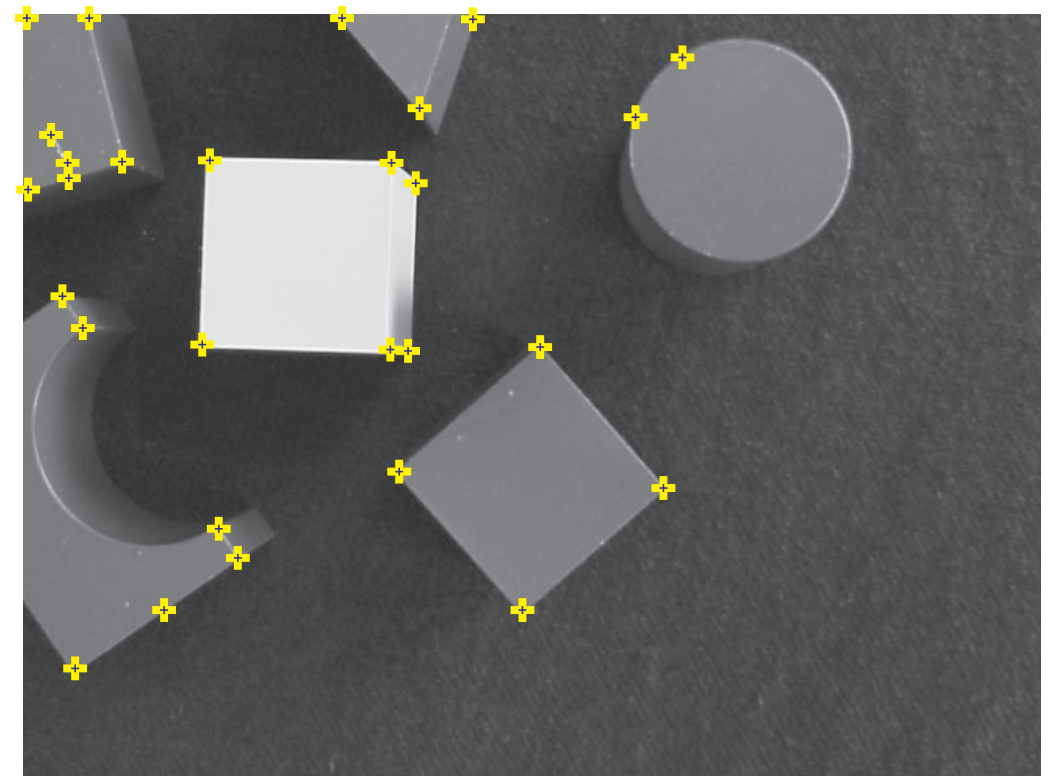
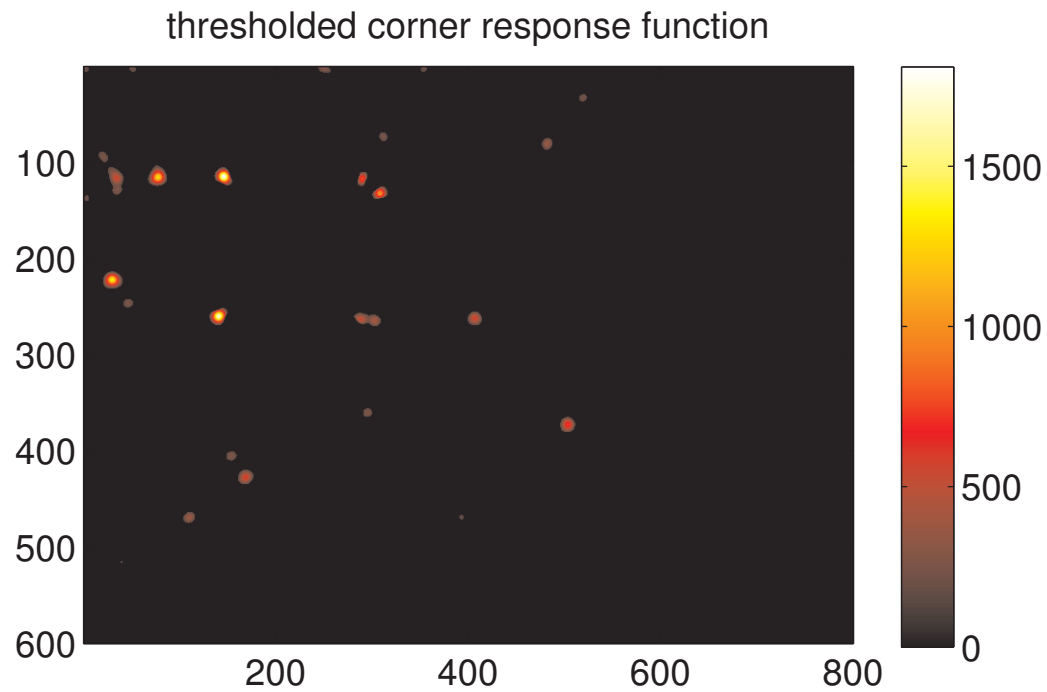
# Corner detection in pictures, (smoothed) derivatives



# Corner detection in pictures, mixed derivatives and reponse function



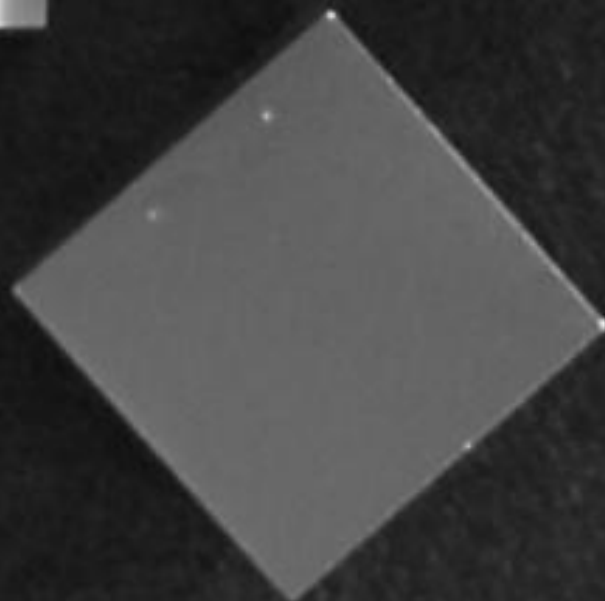
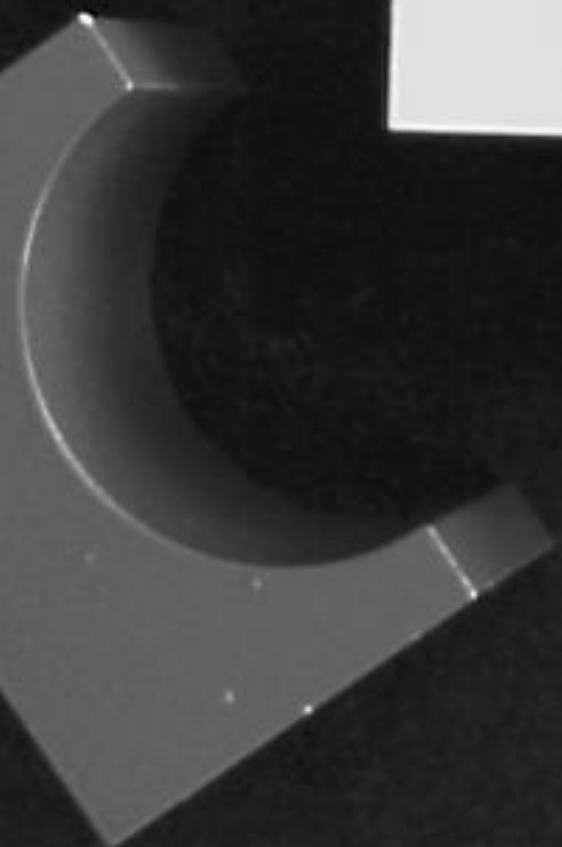
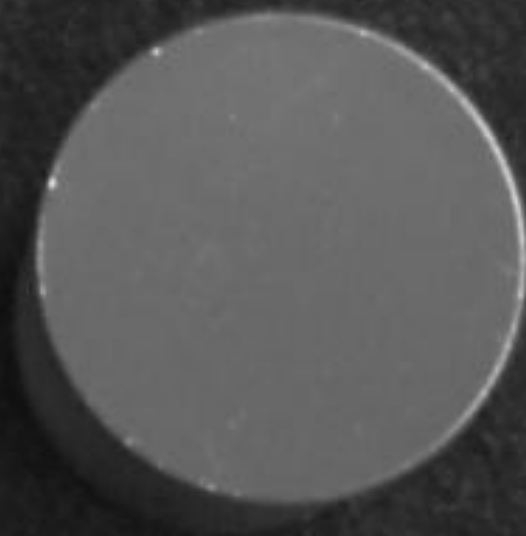
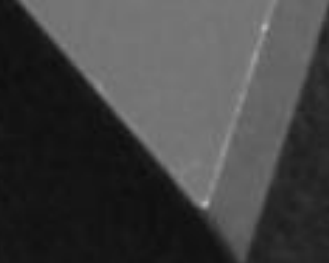
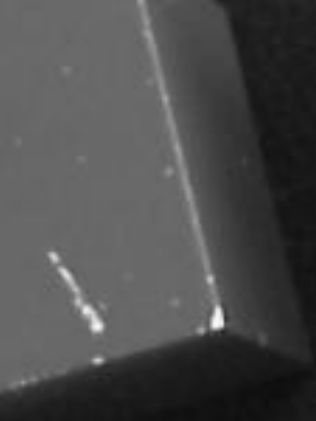
# Corner detection in pictures, thresholding and corners



# References

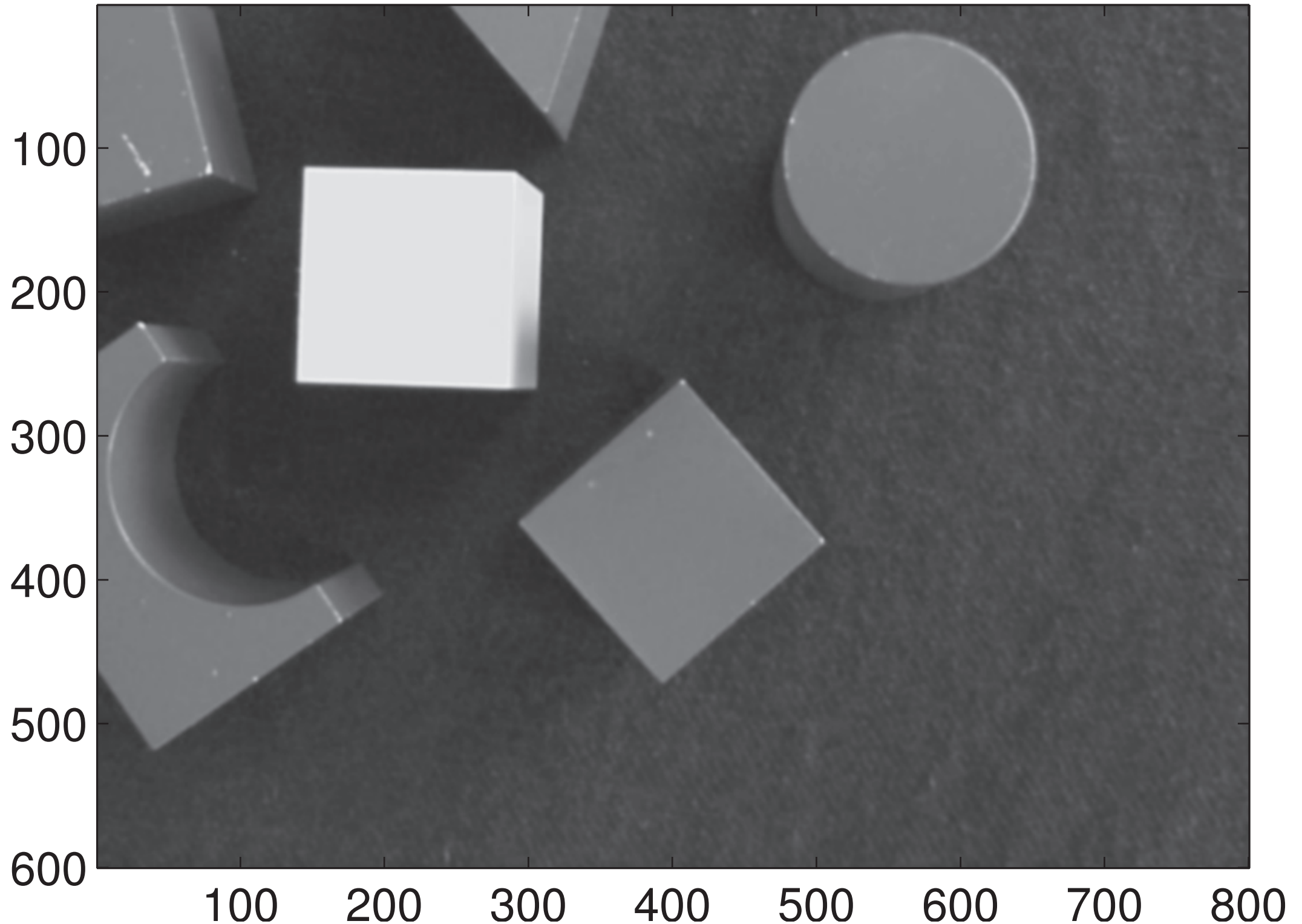
- [1] C. Harris and M. Stephen. A combined corner and edge detection. In M. M. Matthews, editor, *Proceedings of the 4th ALVEY vision conference*, pages 147–151, University of Manchester, England, September 1988. on-line copies available on the web.
- [2] Tomáš Svoboda, Jan Kybic, and Václav Hlaváč. *Image Processing, Analysis and Machine Vision. A MATLAB Companion*. Thomson, 2007. Accompanying www site <http://visionbook.felk.cvut.cz>.

**End**



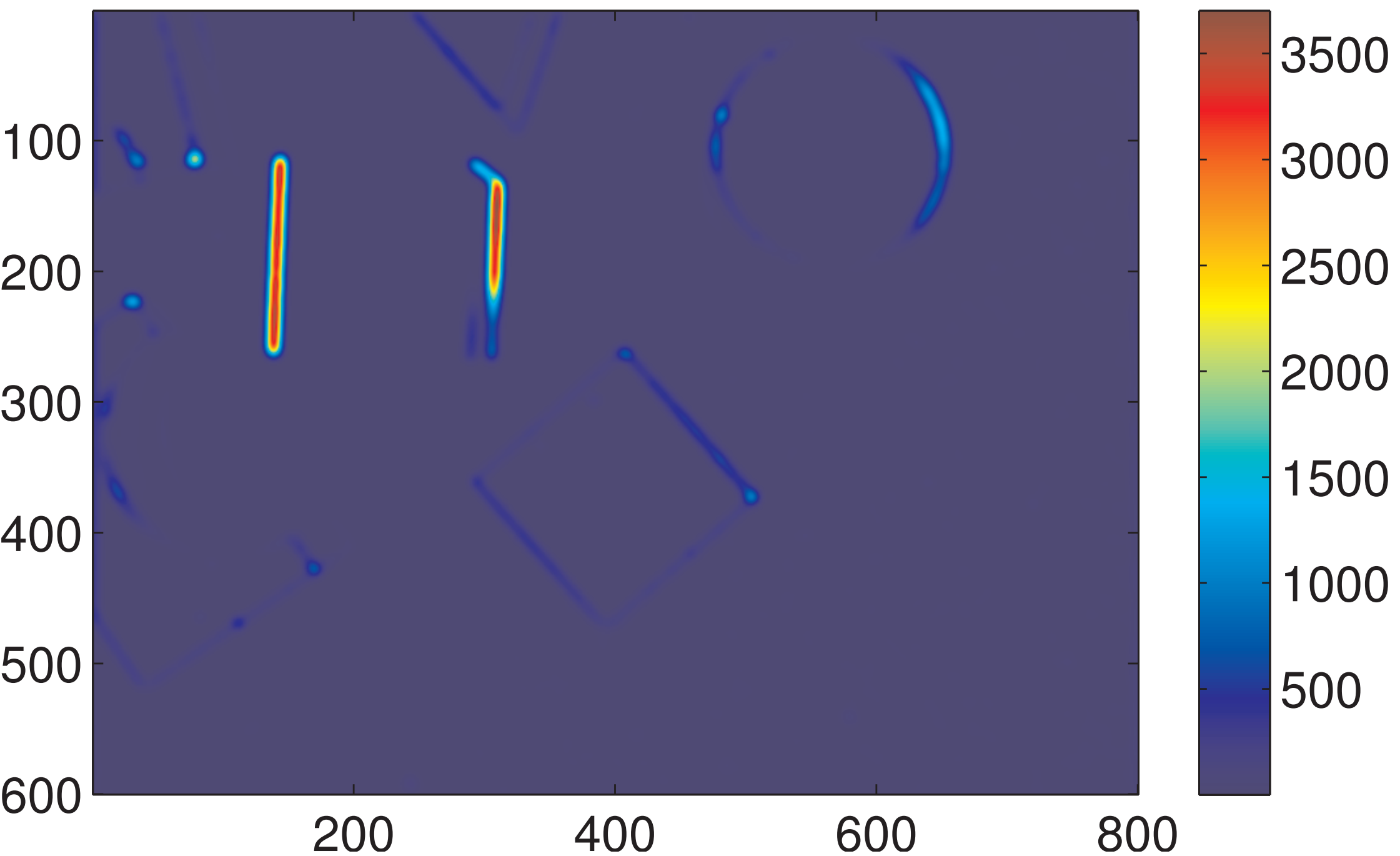


smoothed image

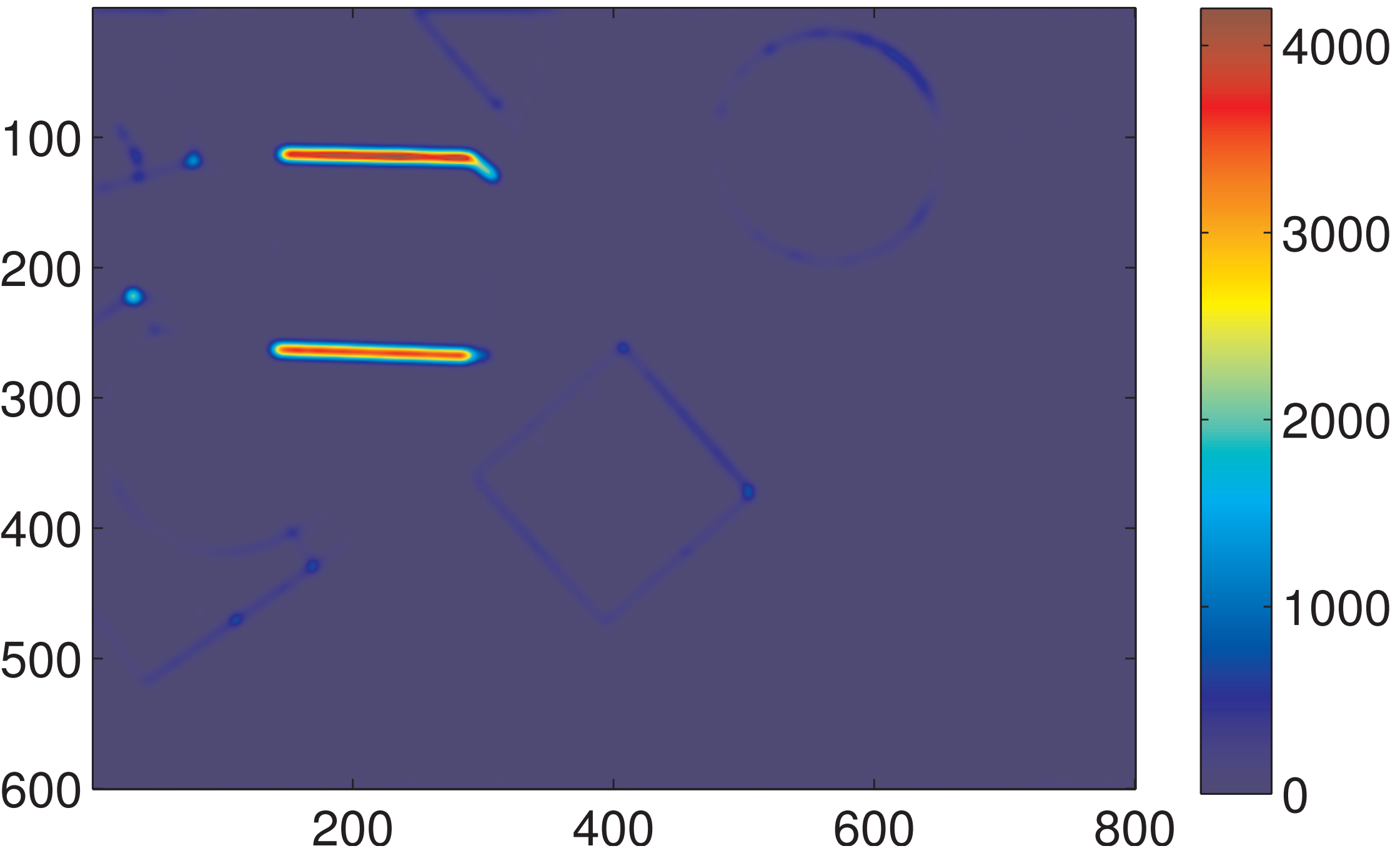


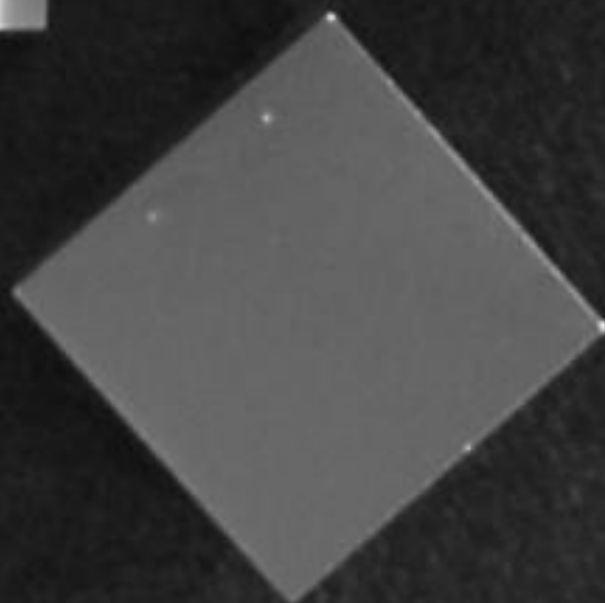
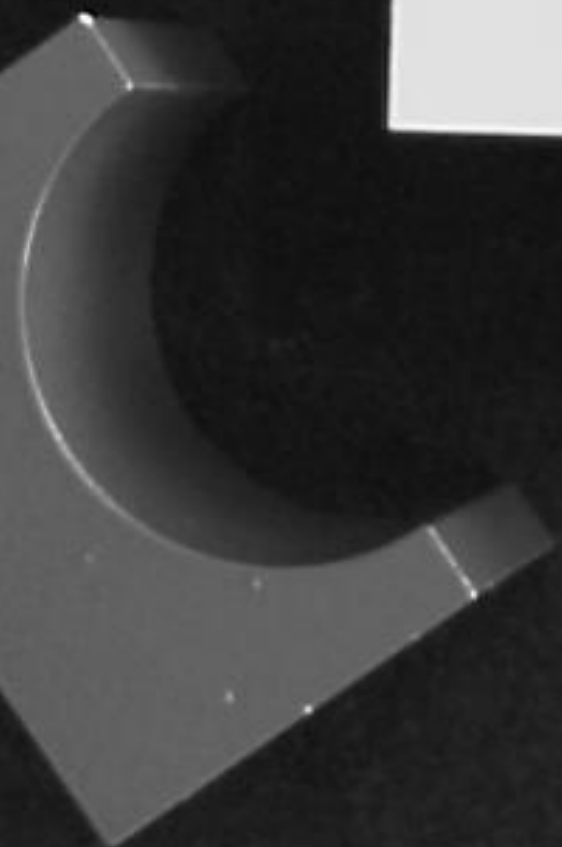
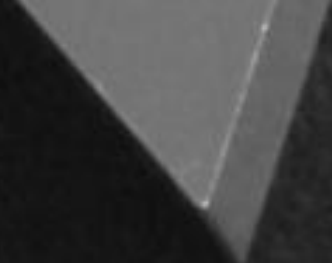


smoothed (x-derivatives)<sup>2</sup>

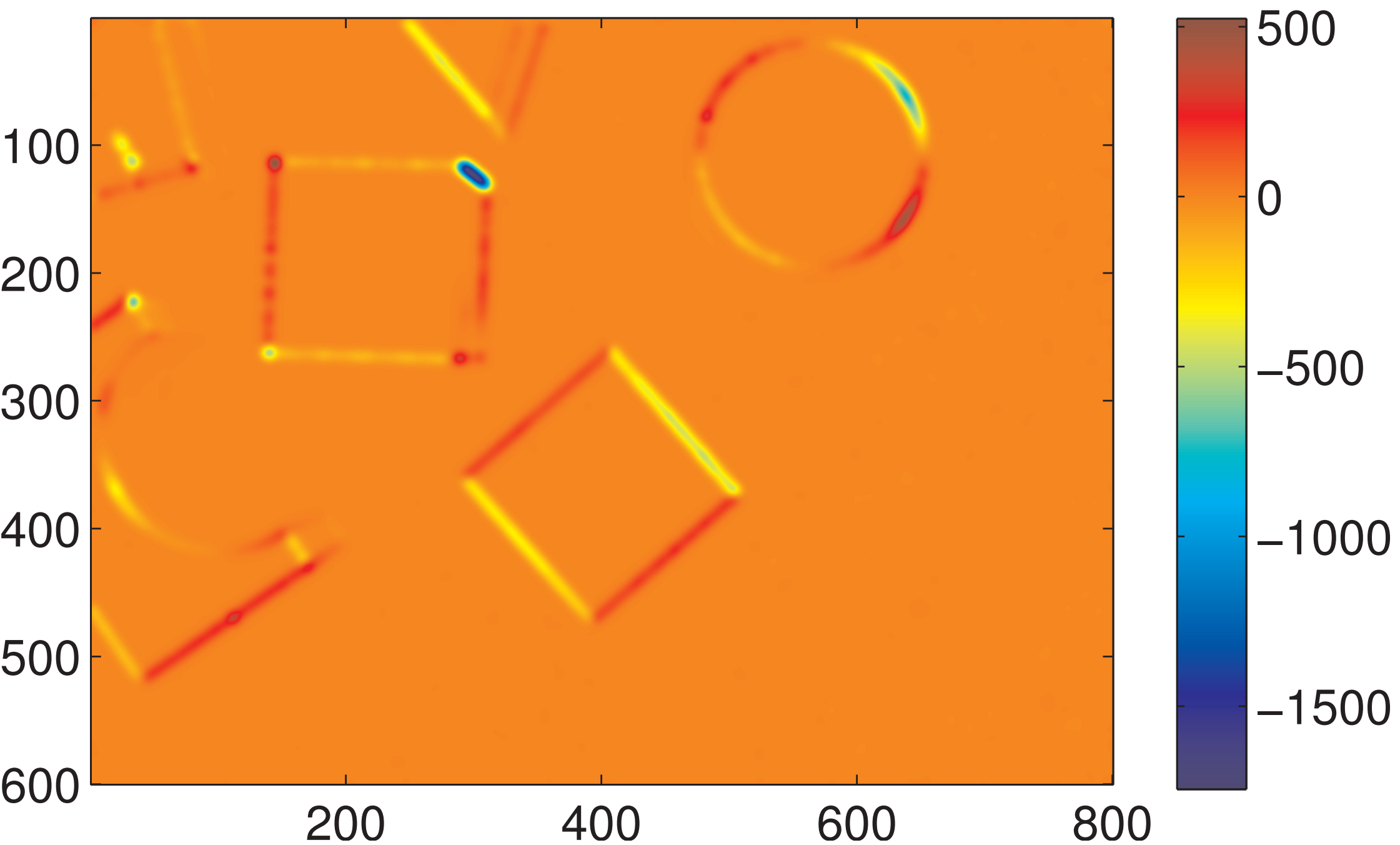


smoothed (y-derivatives)<sup>2</sup>

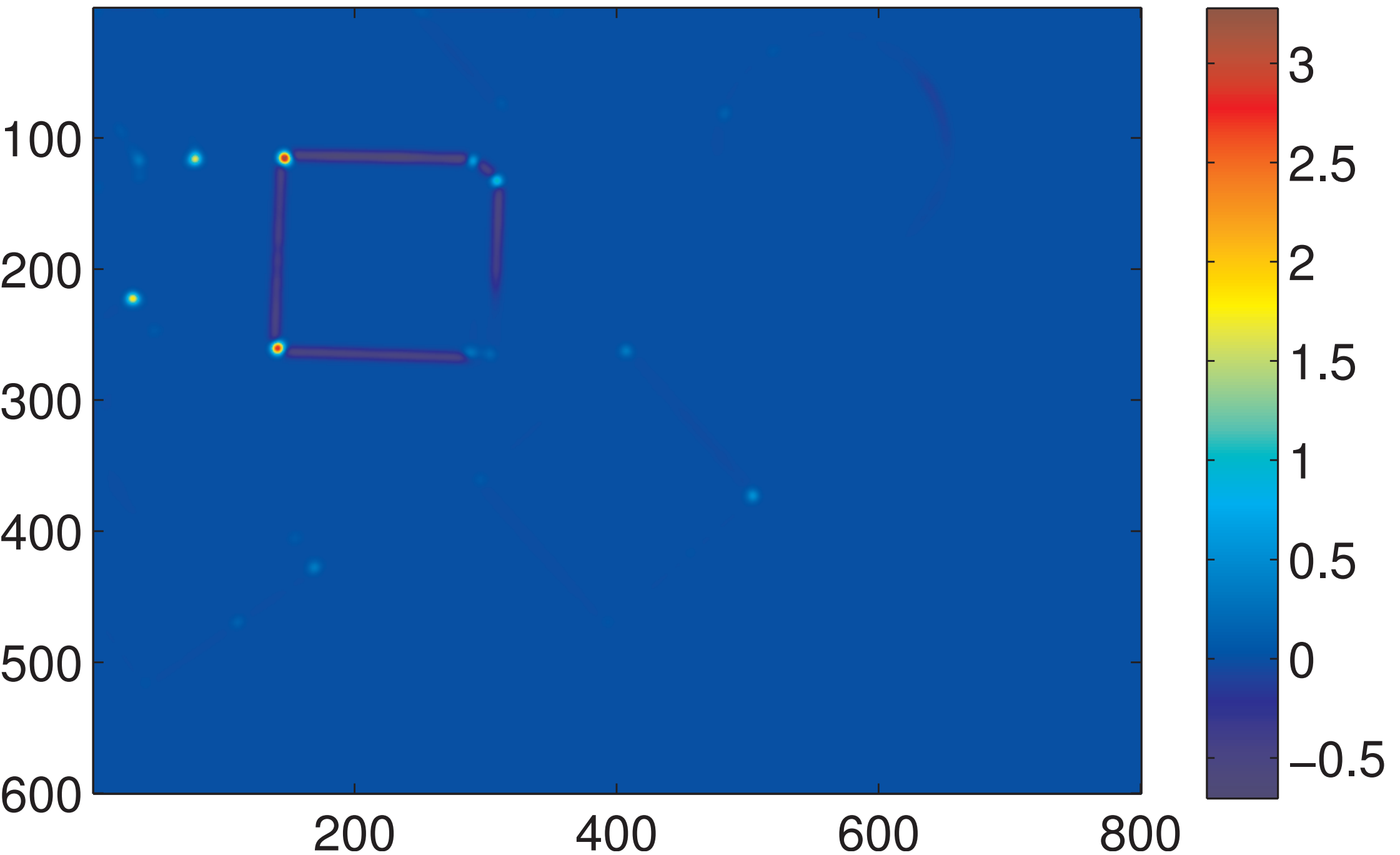


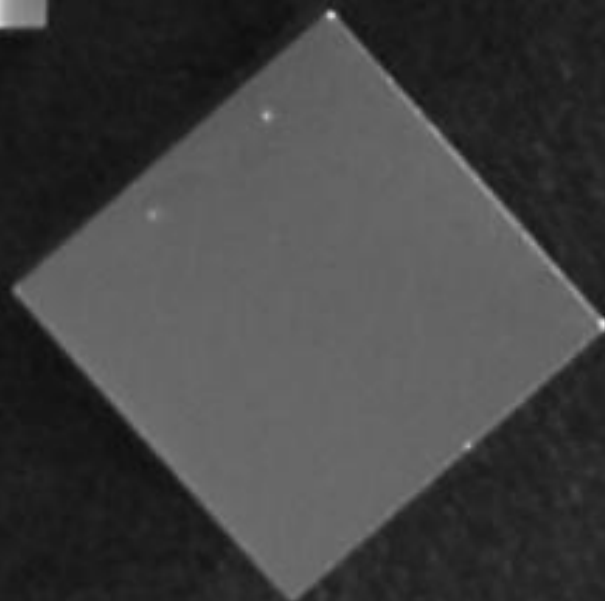
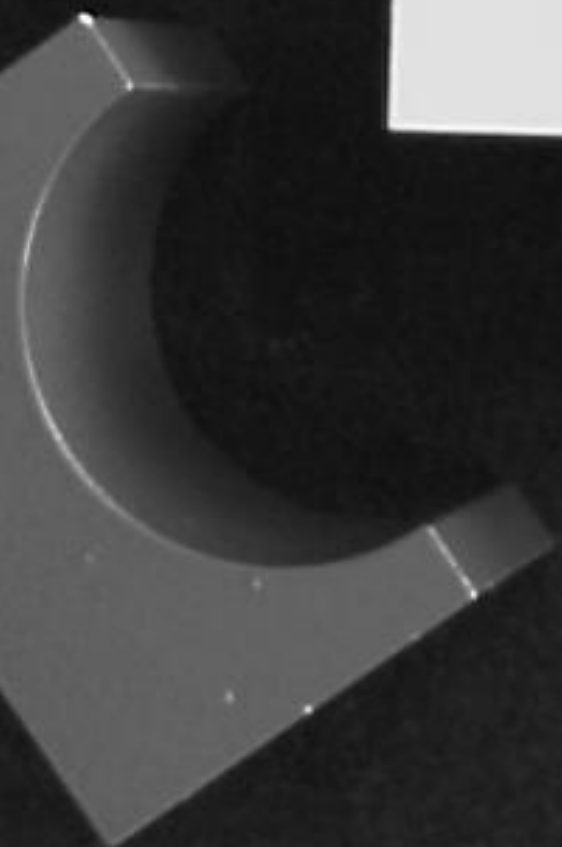
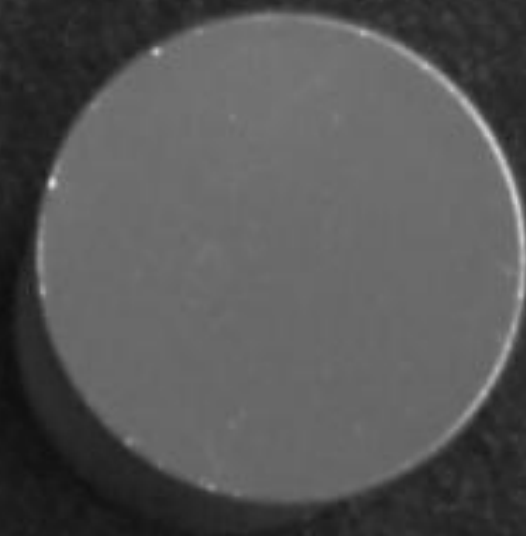
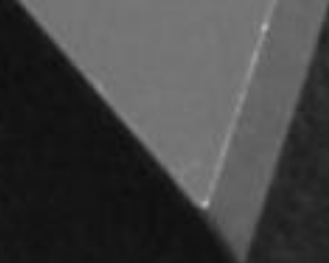


smoothed (x-derivatives).\*(y-derivatives)



corner response function





# thresholded corner response function

