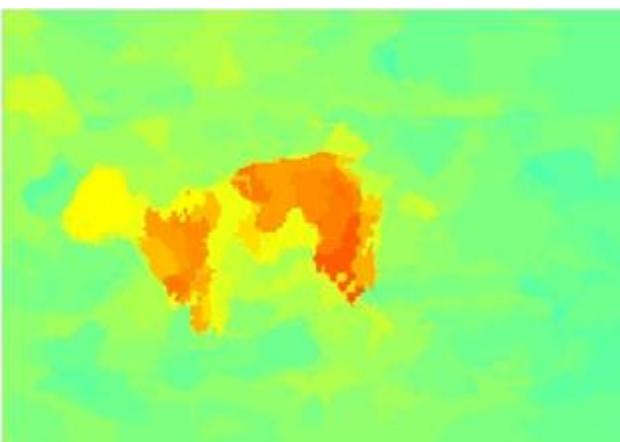


Feature Analysis – Edges and Blobs

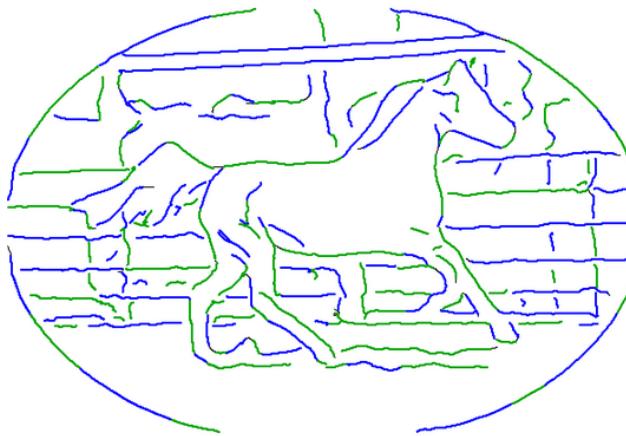
Gonzalez & Woods Digital Image
Processing – Chapter 10

Feature Analysis

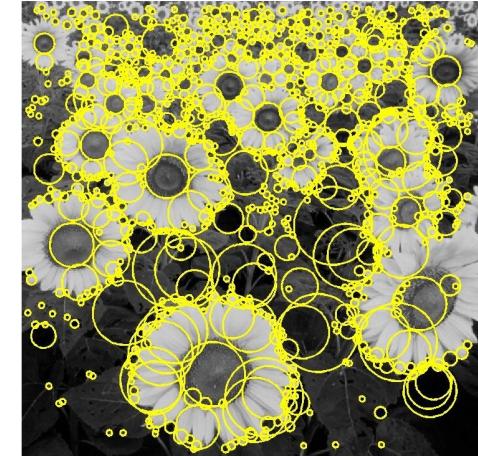
Regions/Segmentation



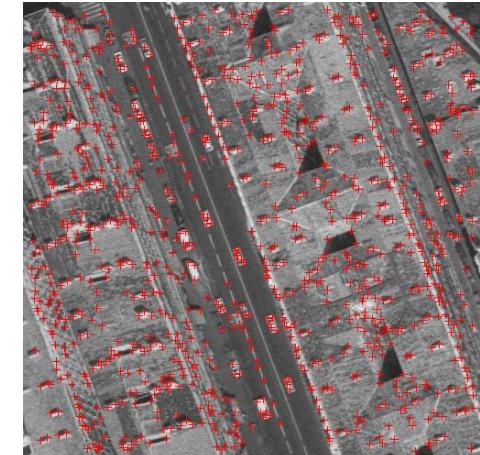
Contours/Line segments



Interest Points
Blobs

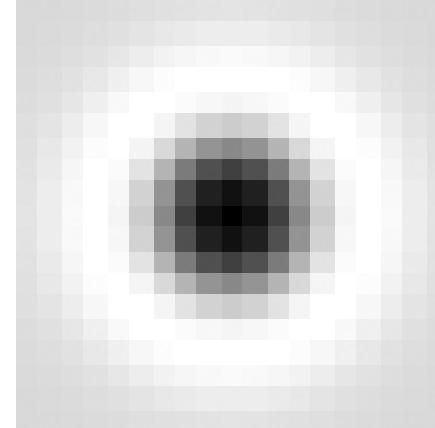
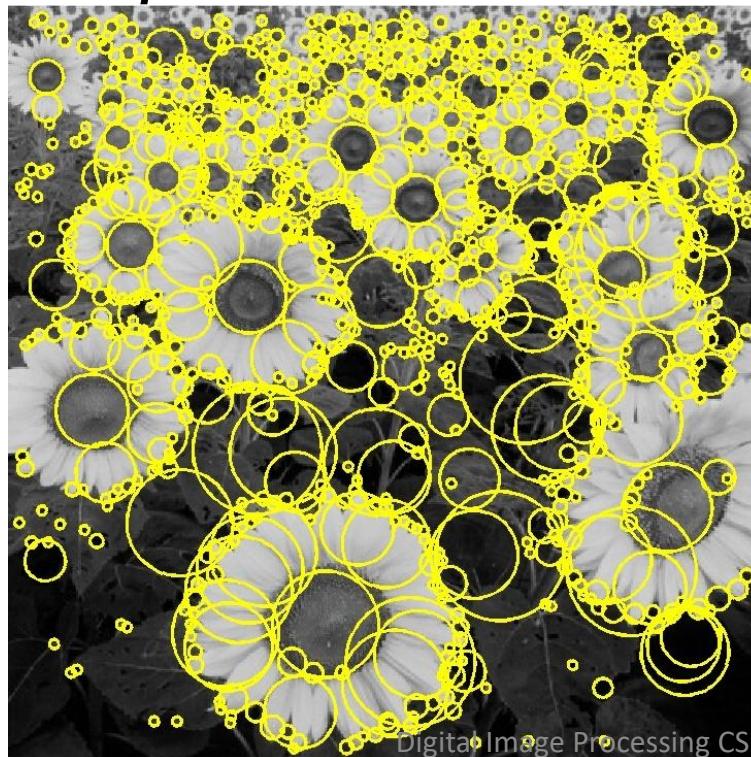


Corners



Blob detection: Basic idea

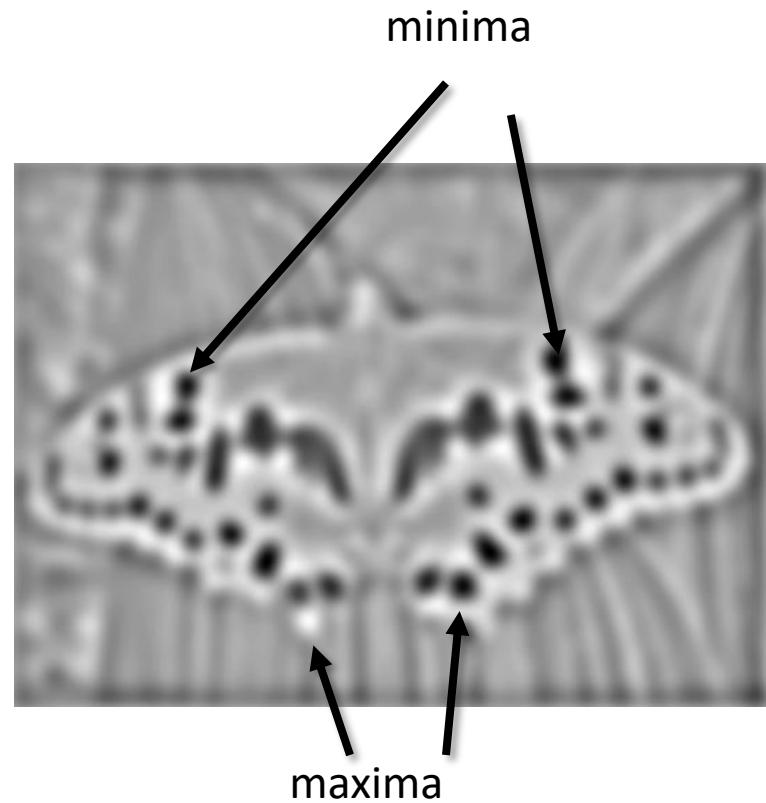
- To detect blobs, convolve the image with a “blob filter” at **multiple scales** and look for **extrema of filter response** in the resulting *scale space*



Blob detection: Basic idea



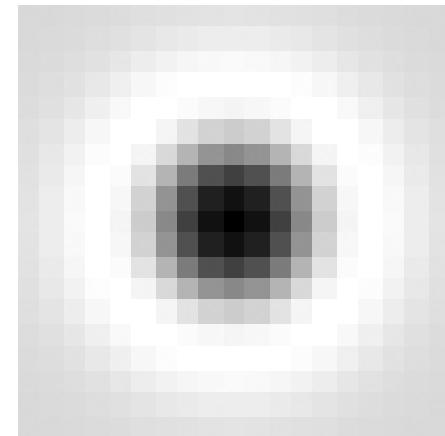
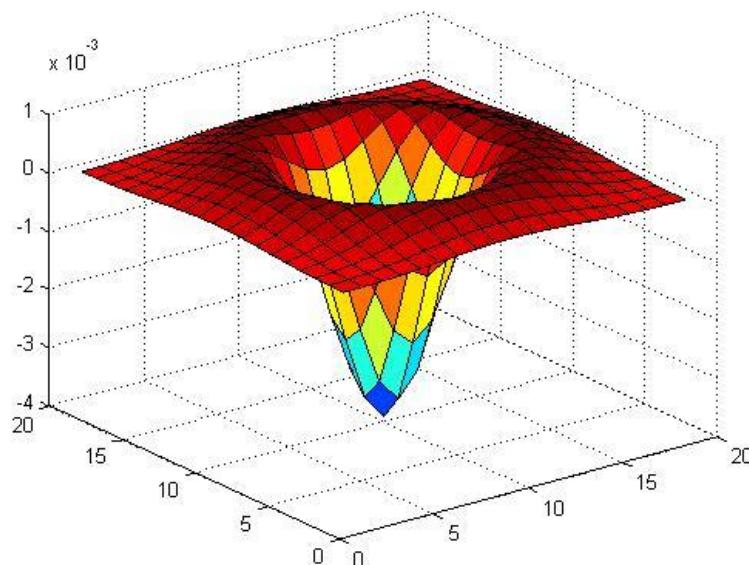
$$* \quad =$$



- Find maxima *and minima* of blob filter response in space *and scale*

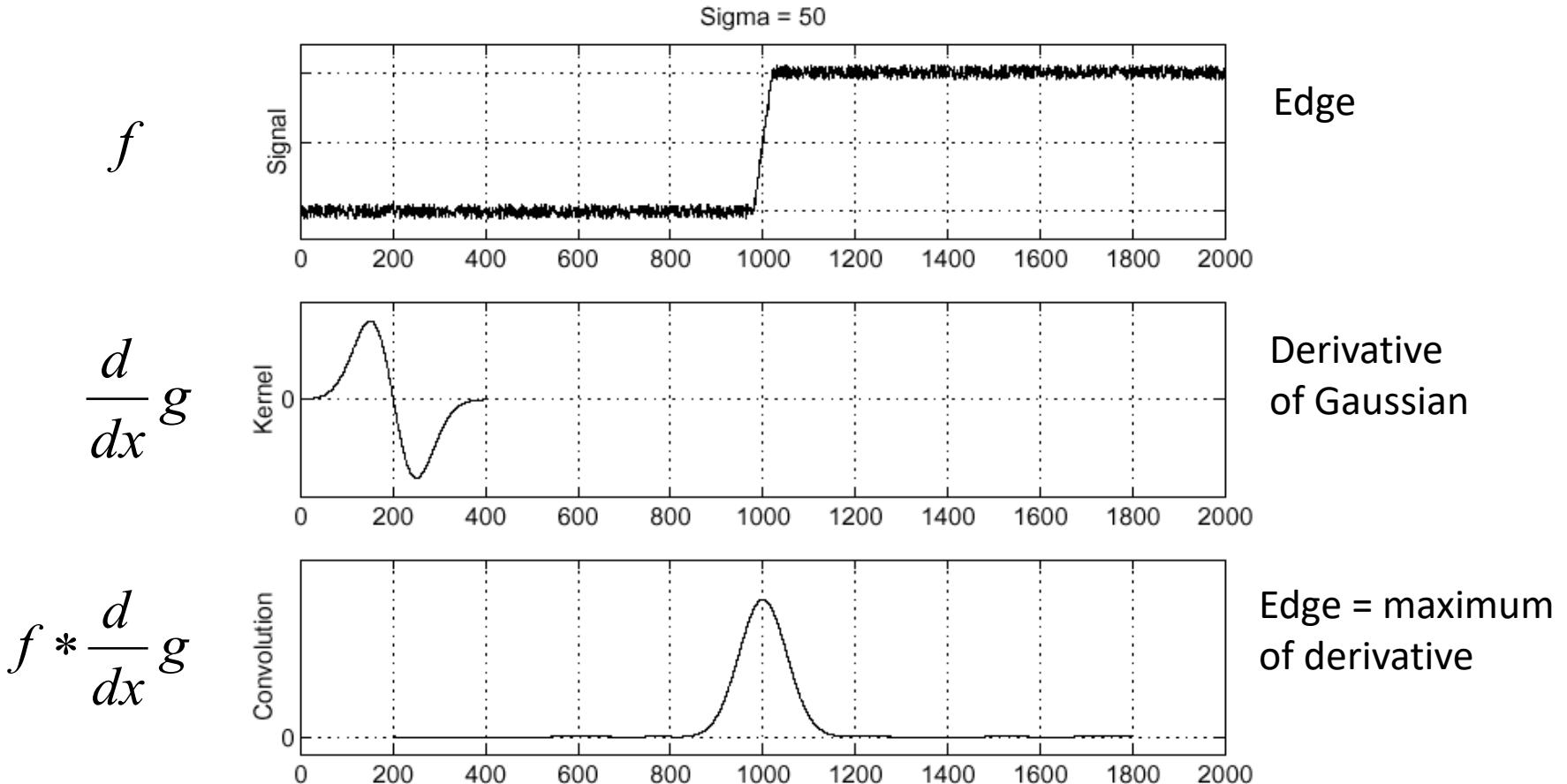
Blob filter

- Laplacian of Gaussian: Circularly symmetric operator for blob detection in 2D



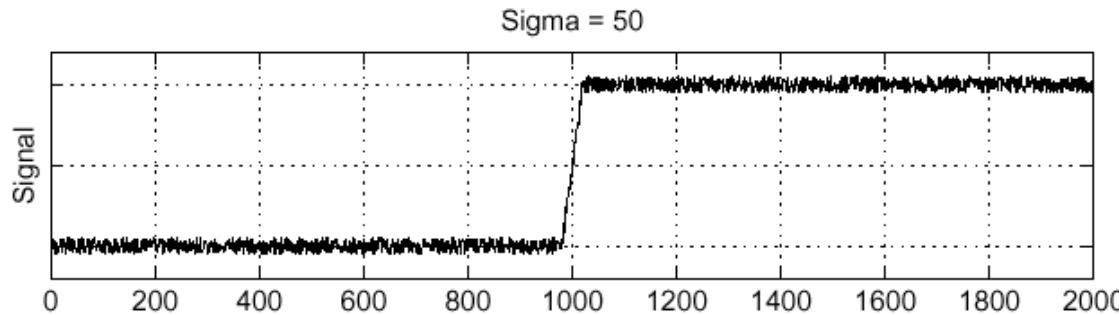
$$\nabla^2 g = \frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2}$$

Recall: Edge detection

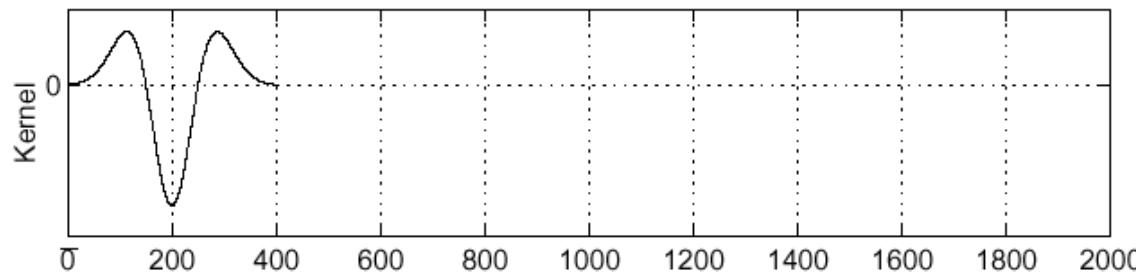


Edge detection, Take 2

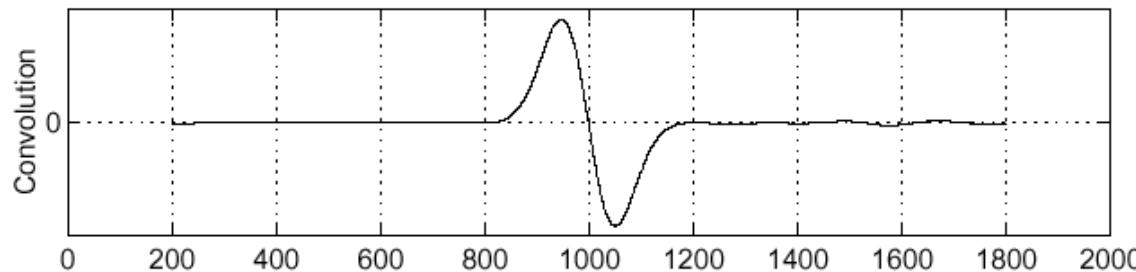
f



$\frac{d^2}{dx^2} g$

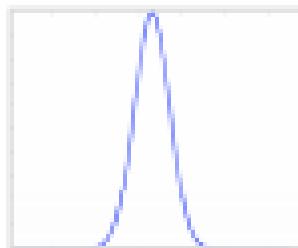


$f * \frac{d^2}{dx^2} g$

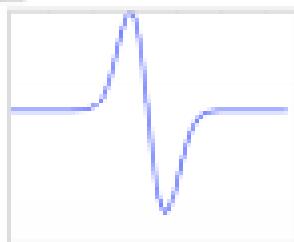


1D Gaussian and Derivatives

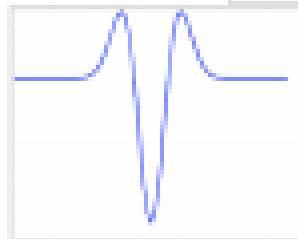
$$g(x) = e^{-\frac{x^2}{2\sigma^2}}$$



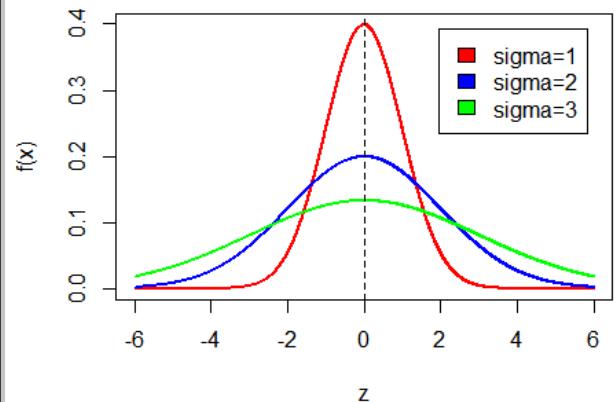
$$g'(x) = -\frac{1}{2\sigma^2} 2xe^{-\frac{x^2}{2\sigma^2}} = -\frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}$$



$$g''(x) = \left(\frac{x^2}{\sigma^4} - \frac{1}{\sigma^2}\right) e^{-\frac{x^2}{2\sigma^2}}$$

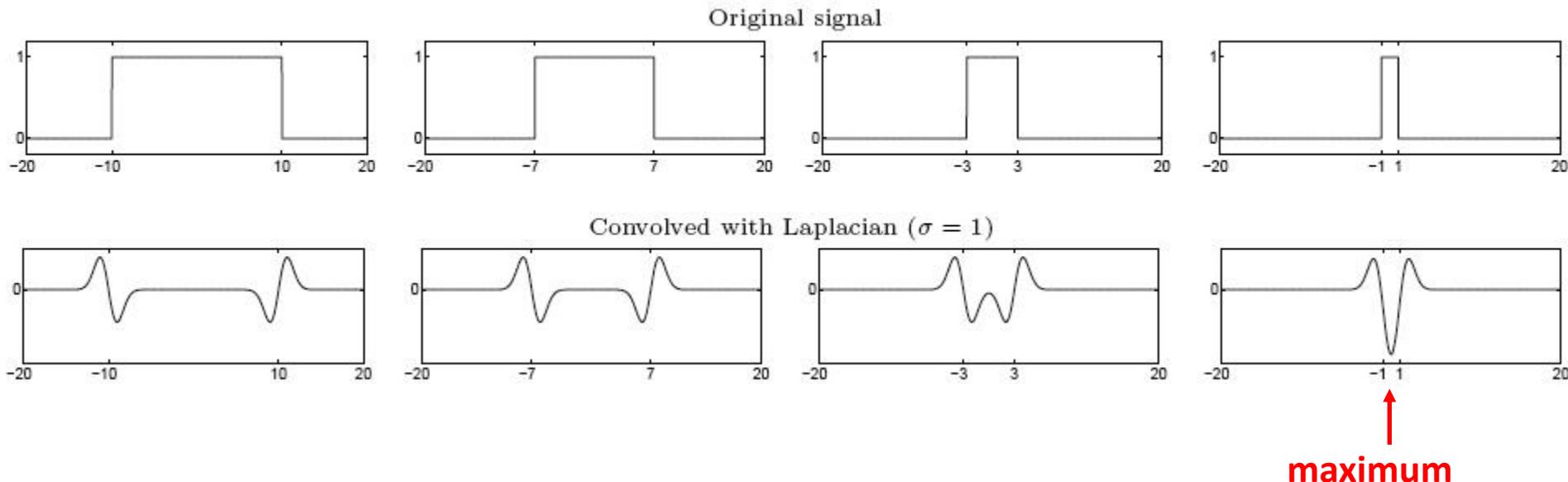
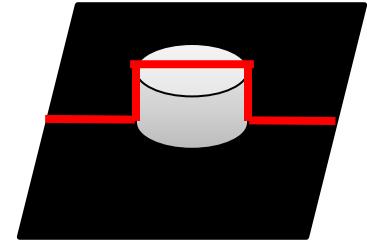


Normal density function with different variance

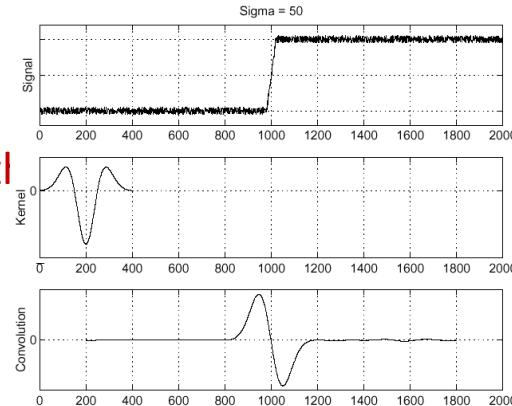


From edges to blobs

- Edge = ripple
- Blob = superposition of two ripples

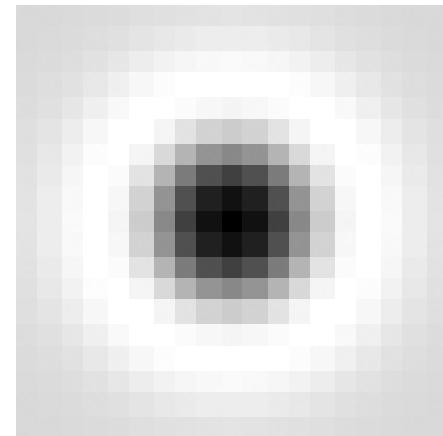
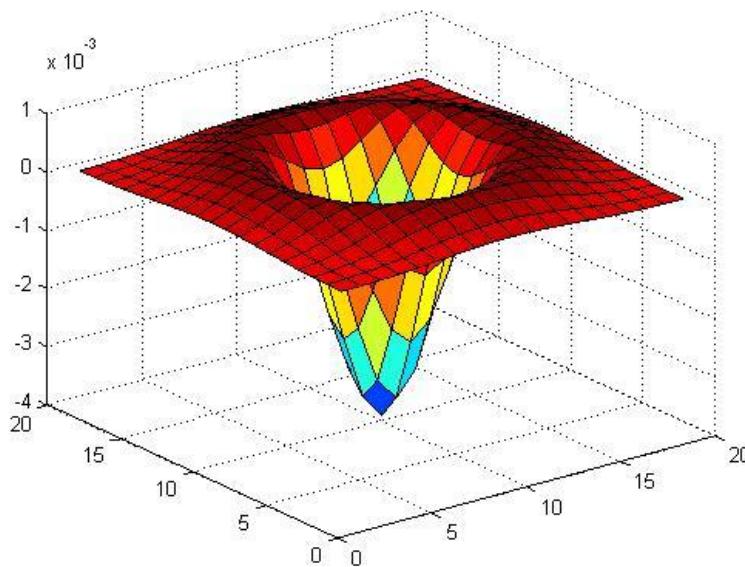


Spatial selection: the magnitude of the Laplacian response will achieve a maximum at the center of the blob, provided the scale of the Laplacian is “matched” to the scale of the edge.



Blob detection in 2D

- Laplacian of Gaussian: Circularly symmetric operator for blob detection in 2D

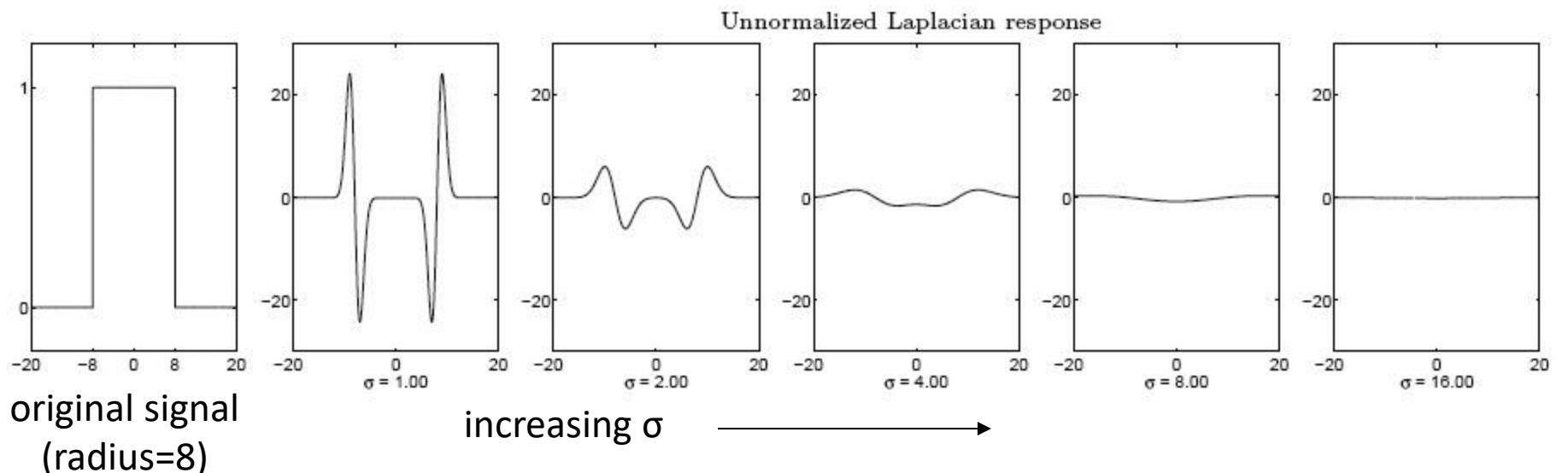


Scale-normalized:

$$\nabla_{\text{norm}}^2 g = \sigma^2 \left(\frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2} \right)$$

Scale selection

- We want to find the **characteristic scale** of the **blob** by convolving it with **Laplacians** at several **scales** and looking for the **maximum response**
- However, Laplacian response decays as scale increases:

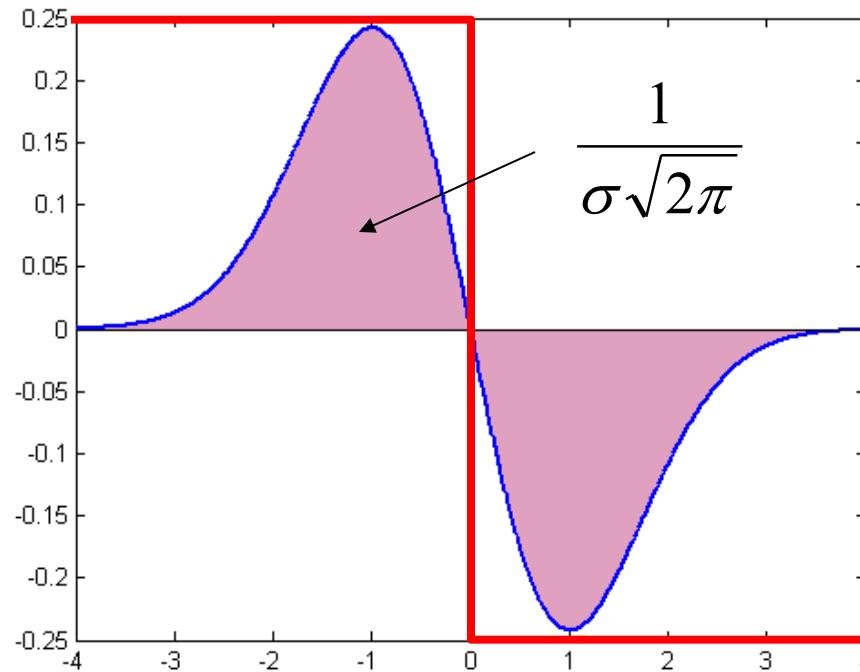


Scale normalization

- The response of a derivative of Gaussian filter to a perfect step edge decreases as σ increases
- To keep response the same (scale-invariant), must multiply Gaussian derivative by σ
- Laplacian is the second Gaussian derivative, so it must be multiplied by σ^2

Scale normalization

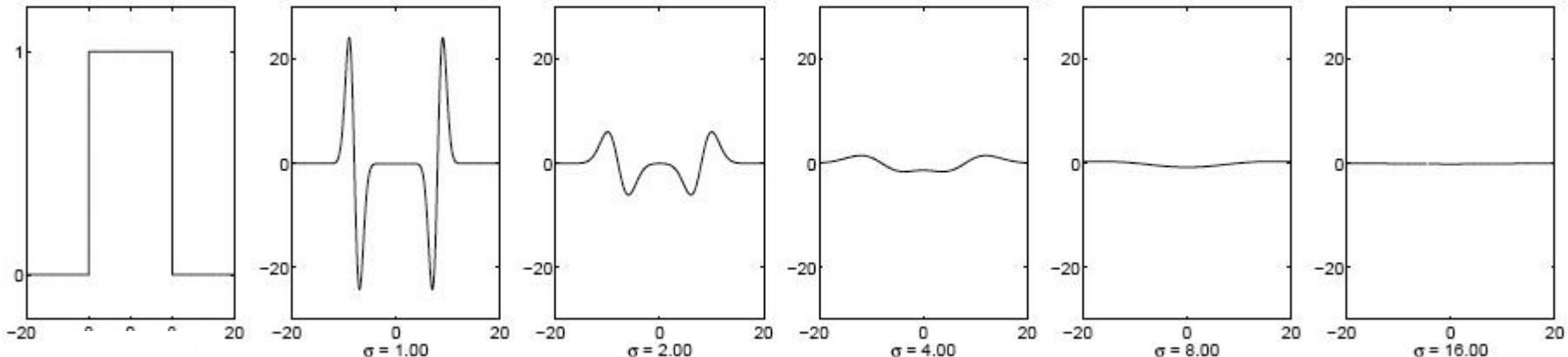
- The response of a derivative of Gaussian filter to a perfect step edge decreases as σ increases



Effect of scale normalization

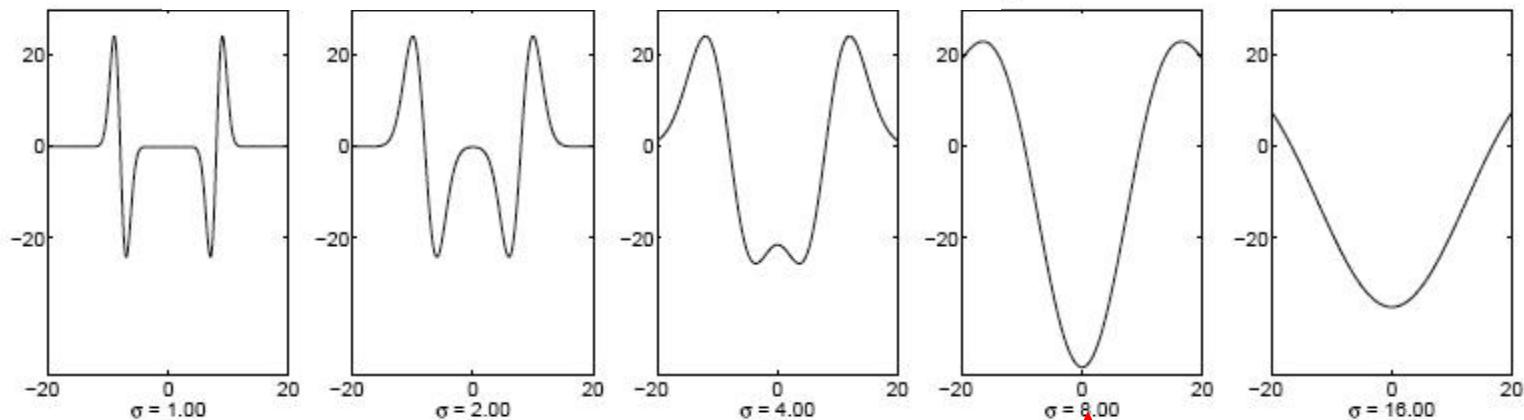
$$\nabla^2 g = \frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2}$$

Unnormalized Laplacian response



$$\nabla_{\text{norm}}^2 g = \sigma^2 \left(\frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2} \right)$$

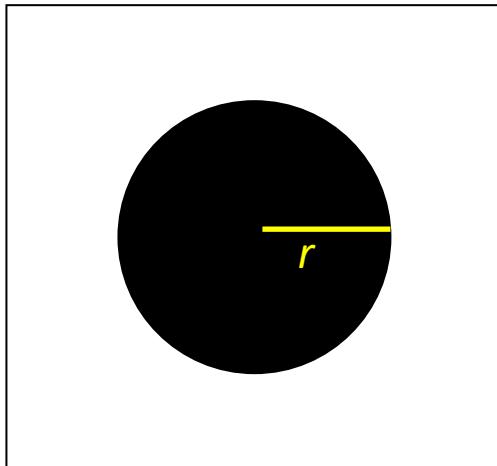
Scale-normalized Laplacian response



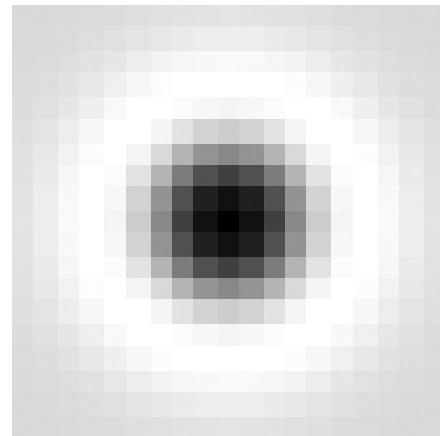
maximum

Scale selection

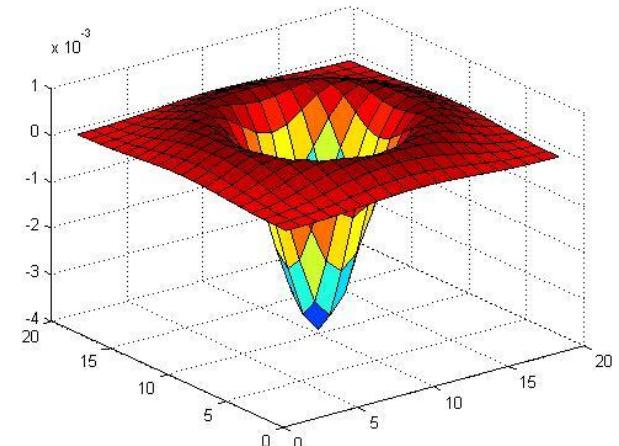
- At what scale does the Laplacian achieve a maximum response to a binary circle of radius r ?



image



Laplacian



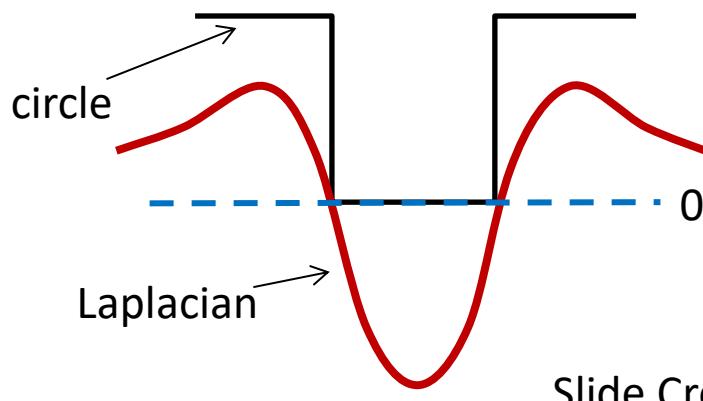
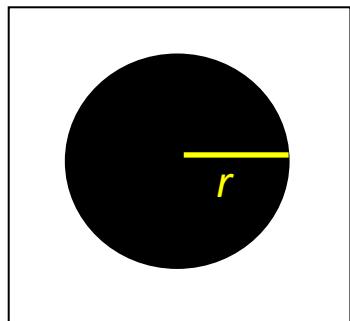
$$(x^2 + y^2 - 2\sigma^2)e^{-(x^2+y^2)/2\sigma^2}$$

Scale selection

- At what scale does the Laplacian achieve a maximum response to a binary circle of radius r ?
- To get maximum response, the zeros of the Laplacian have to be aligned with the circle
- The Laplacian is given by (up to scale): $(x^2 + y^2 - 2\sigma^2)e^{-(x^2+y^2)/2\sigma^2}$
- Therefore, the maximum response occurs at

$$0 = (r^2 - 2\sigma^2)e^{-\frac{r^2}{2\sigma^2}}$$
$$0 = (r^2 - 2\sigma^2)$$

$$\sigma = r / \sqrt{2}.$$

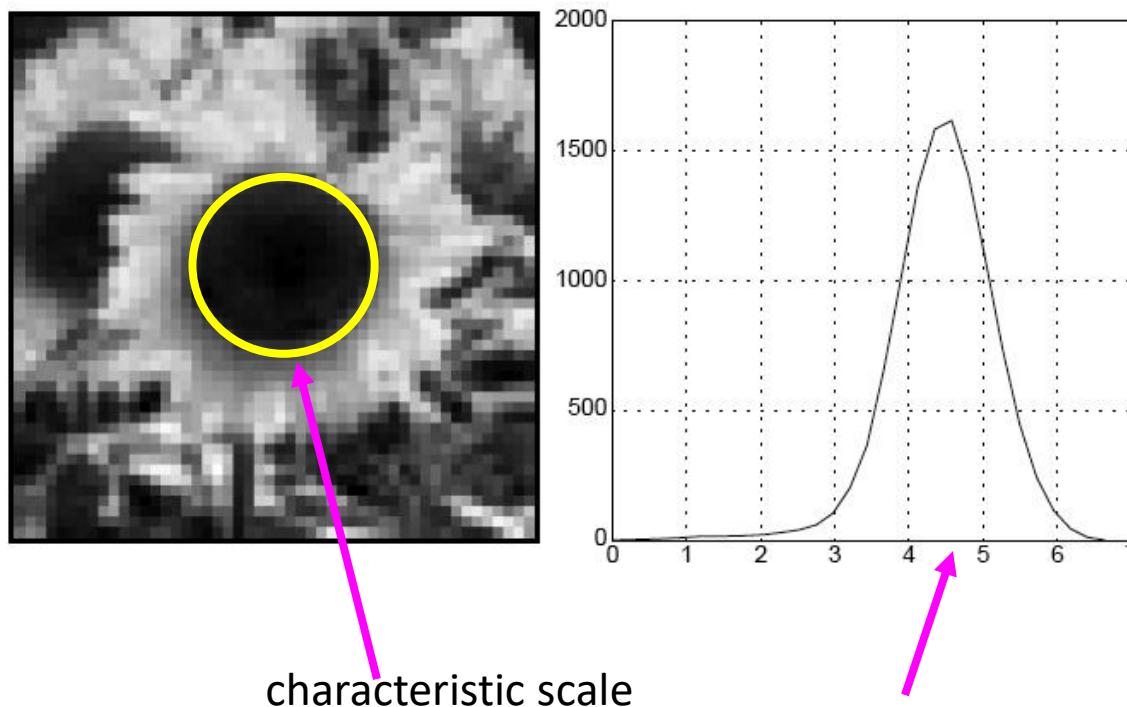


image

Slide Credit: Lazebnik

Characteristic scale

- We define the **characteristic scale of a blob** as the **scale that produces peak of Laplacian response** in the blob center



T. Lindeberg (1998). ["Feature detection with automatic scale selection."](#)
International Journal of Computer Vision **30** (2); pp 77--116.

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Slide Credit: Dr. Svetlana Lazebnik

Scale-space blob detector

1. Convolve image with scale-normalized Laplacian at several scales

Scale-space blob detector: Example



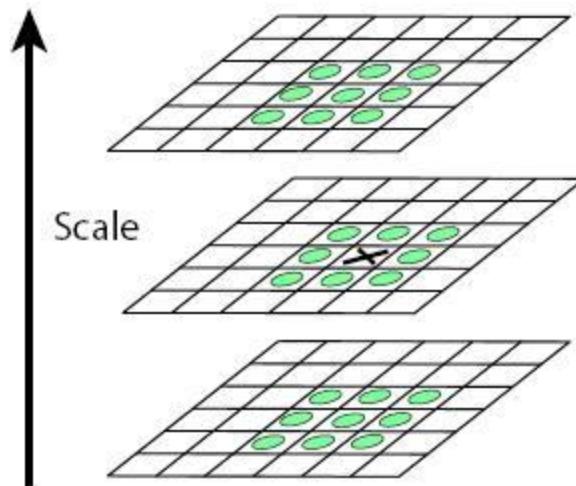
Scale-space blob detector: Example



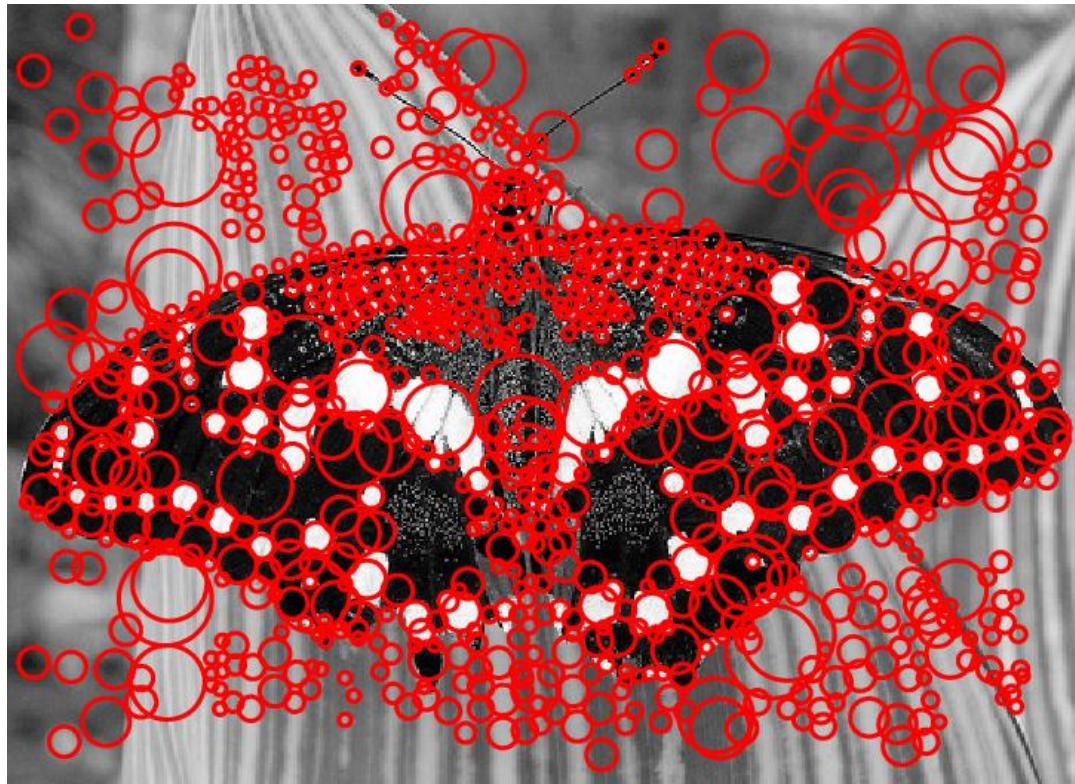
$\sigma = 11.9912$

Scale-space blob detector

1. Convolve image with scale-normalized Laplacian at several scales
2. Find maxima of squared Laplacian response in scale-space



Scale-space blob detector: Example



Efficient implementation

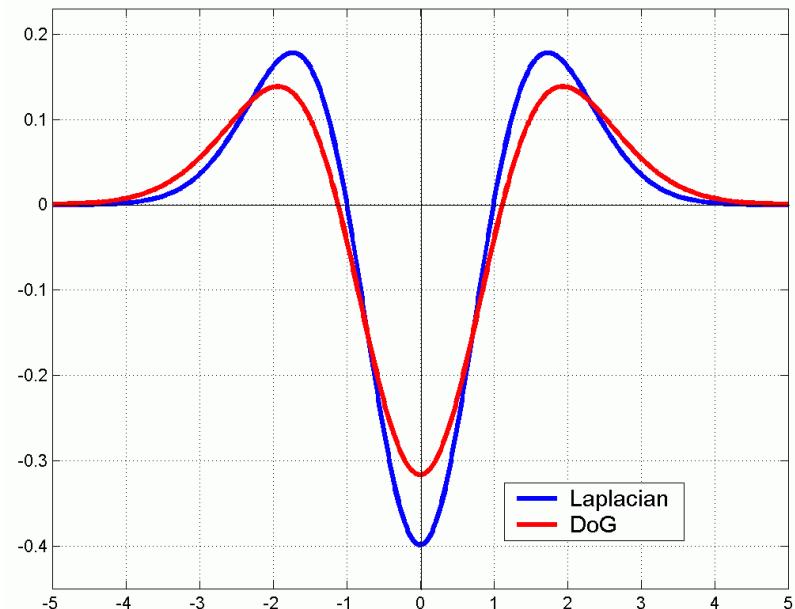
- Approximating the Laplacian with a difference of Gaussians:

$$L = \sigma^2 \left(G_{xx}(x, y, \sigma) + G_{yy}(x, y, \sigma) \right)$$

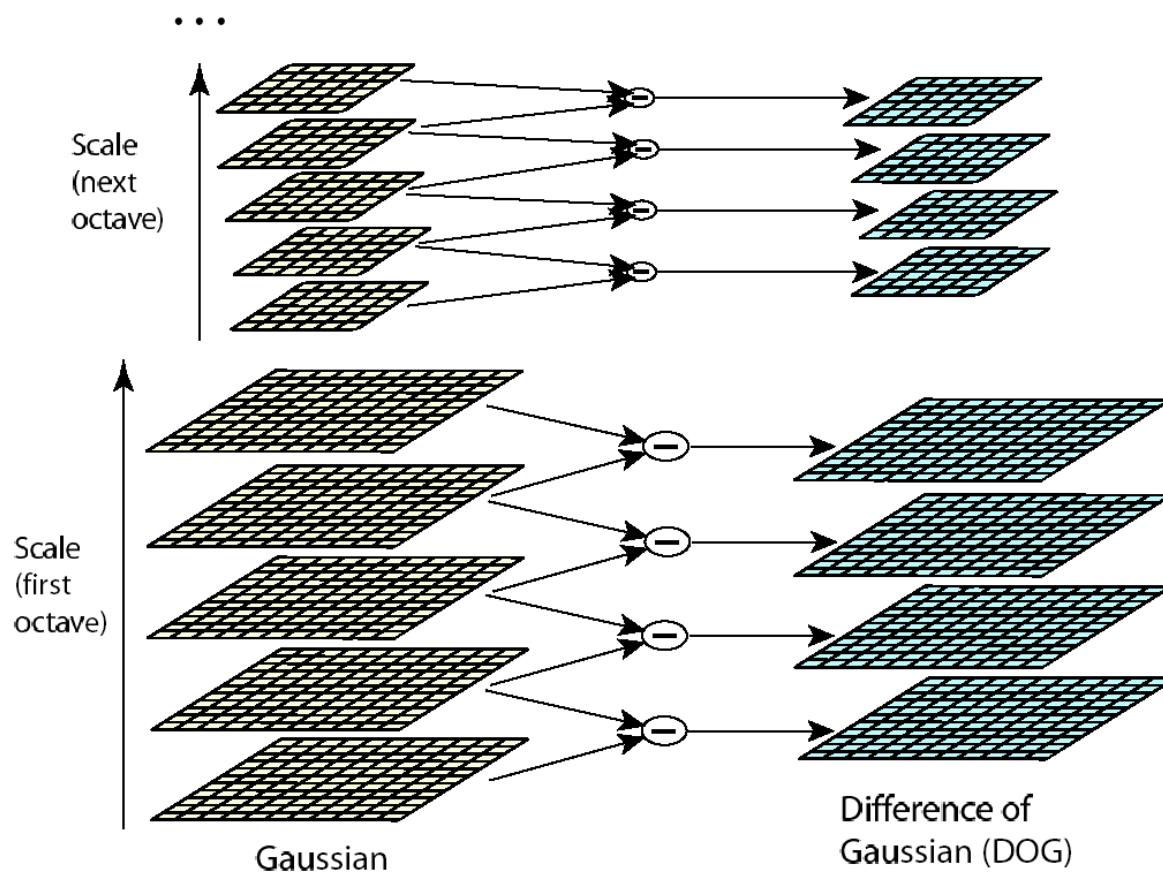
(Laplacian)

$$DoG = G(x, y, k\sigma) - G(x, y, \sigma)$$

(Difference of Gaussians)

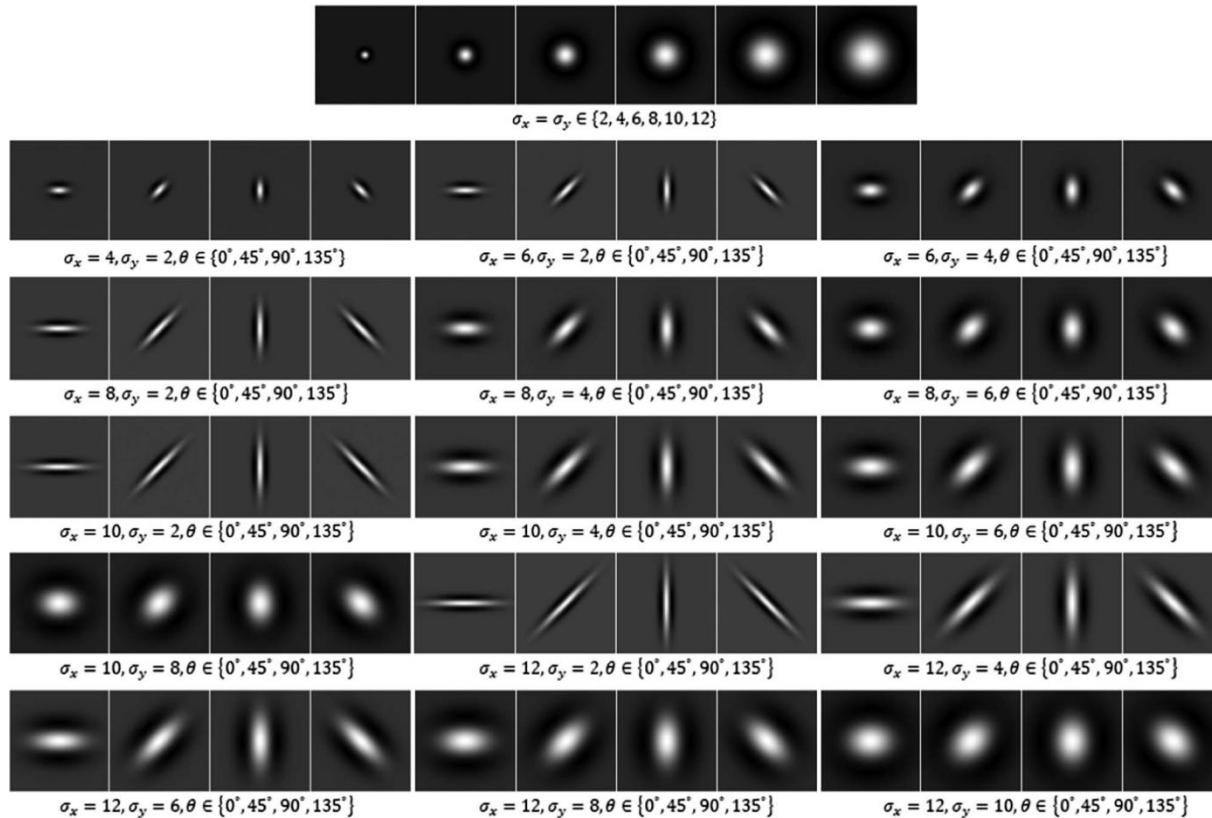


Efficient implementation

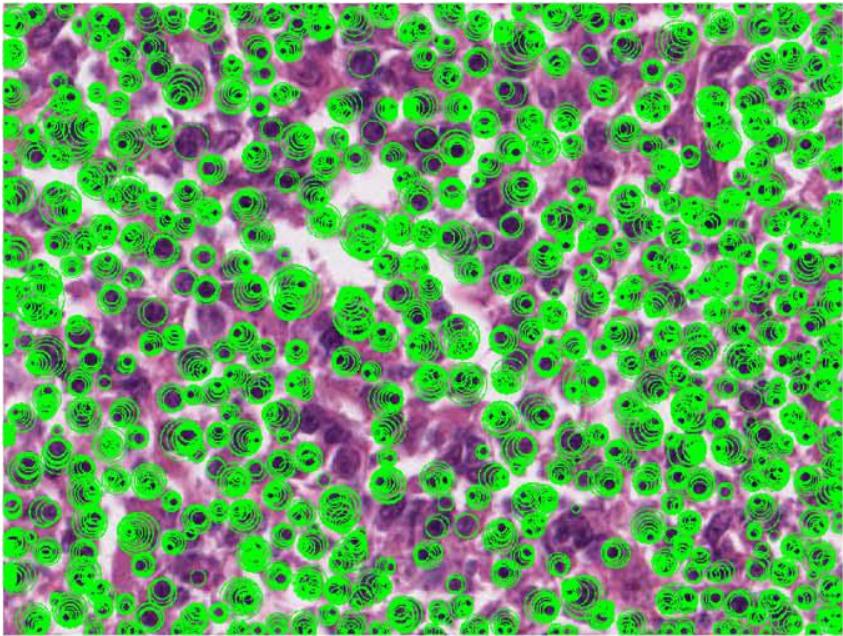


David G. Lowe. "[Distinctive image features from scale-invariant keypoints.](#)" IJCV 60 (2), pp. 91-110, 2004.

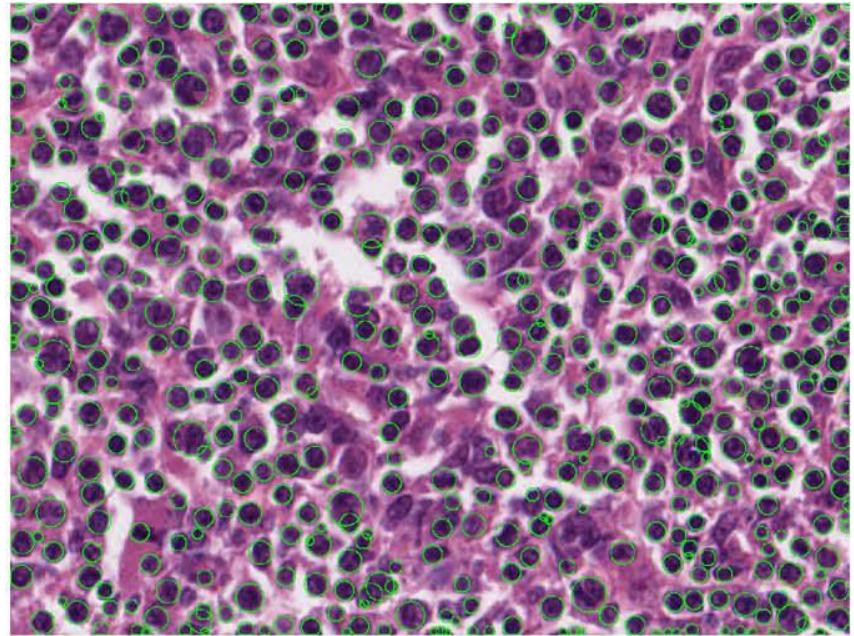
Generalized Laplacian of Gaussian filter



Kong, Hui, Hatice Cinar Akakin, and Sanjay E. Sarma. "A generalized Laplacian of Gaussian filter for blob detection and its applications." *IEEE transactions on cybernetics* 43.6 (2013): 1719-1733.



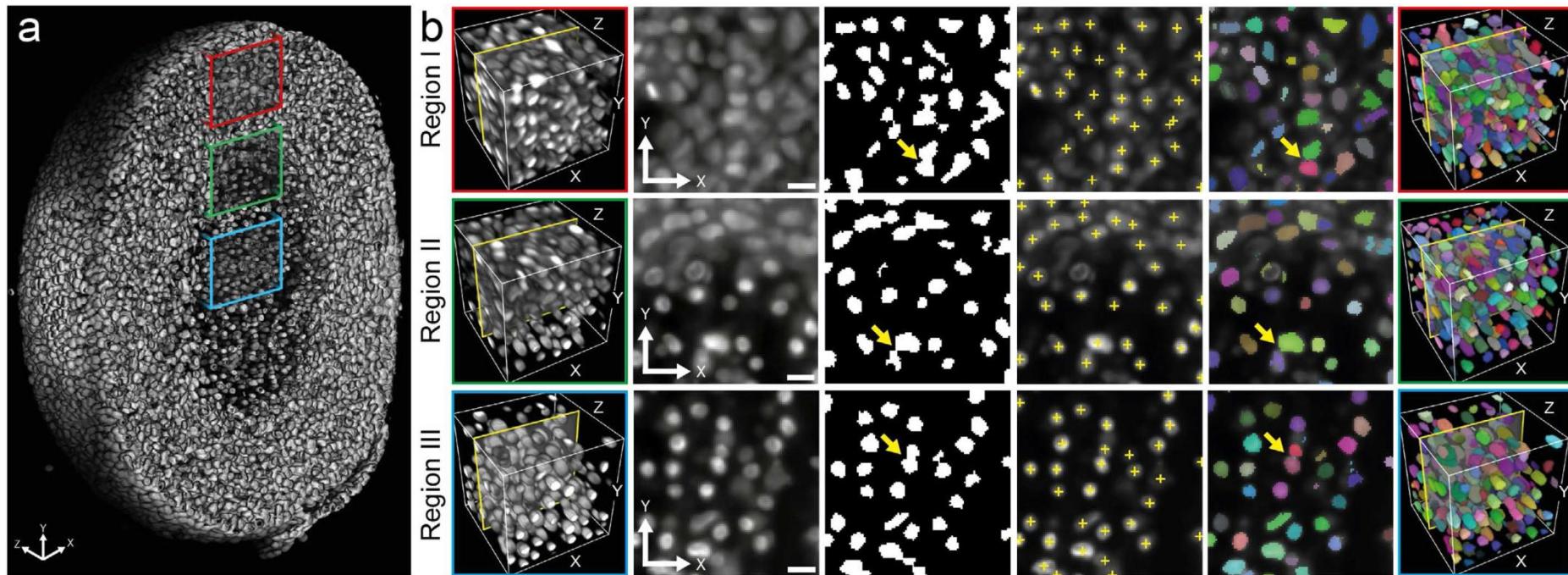
(a)



(b)

(a) and (b) Detected blobs by circular LoG blob detector before and after pruning, respectively.

Kong, Hui, Hatice Cinar Akakin, and Sanjay E. Sarma. "A generalized Laplacian of Gaussian filter for blob detection and its applications." *IEEE transactions on cybernetics* 43.6 (2013): 1719-1733.



Multiscale image analysis reveals structural heterogeneity of the cell microenvironment in homotypic spheroids

Alexander Schmitz

, Sabine C. Fischer

, Christian Mattheyer

, Francesco Pampaloni

& Ernst H. K. Stelzer

Scientific Reports 7, Article number: 43693 Digital Image Processing CS 4650/7650

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Dataset S9

