

### 2D Gabor functions and filters for image processing and computer vision



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Most of the images in this presentation were generated with the on-line simulation programs available at:

http://matlabserver.cs.rug.nl

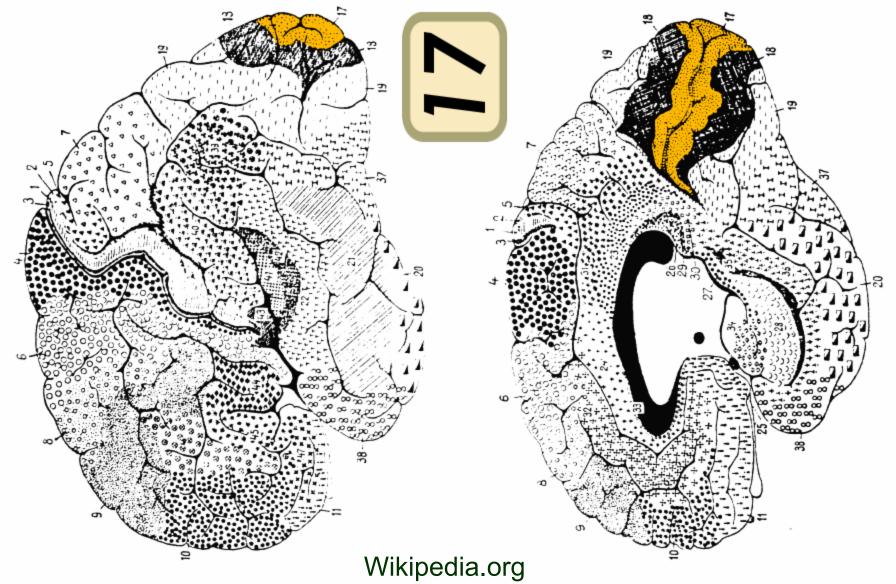


## Neurophysiologic background



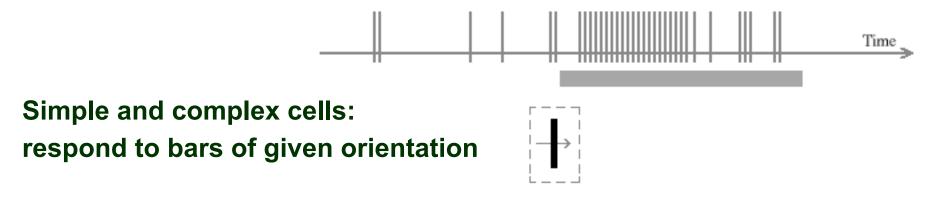
#### Primary visual cortex (striate cortex or V1)

#### **Brodmann area 17**





#### References to origin



D.H. Hubel and T.N. Wiesel: Receptive fields, binocular interaction and functional architecture in the cat's visual cortex, Journal of Physiology (London), vol. 160, pp. 106--154, 1962.

D.H. Hubel and T.N. Wiesel: Sequence regularity and geometry of orientation columns in the monkey striate cortex, Journal of Computational Neurology, vol. 158, pp. 267--293, 1974.

D.H. Hubel: Exploration of the primary visual cortex, 1955-78, Nature, vol. 299, pp. 515--524, 1982.



# Simple cells and Gabor filters

(or a Platonic view of reality)

Hubel and Wiesel named one type of cell "simple" because they shared the following properties:

- Their receptive fields have distinct excitatory and inhibitory regions.
- These regions follow the summation property.
- These regions have mutual antagonism excitatory and inhibitory regions balance themselves out in diffuse lighting.
- It is possible to predict responses to stimuli given the map of excitatory and inhibitory regions.

*In engineering terms*:

a simple cell can be characterized by an impulse response.

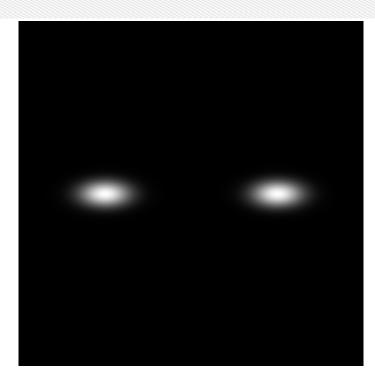


#### Receptive field profiles of simple cells





Frequency domain



#### How are they determined?

- recording responses to bars
- recording responses to gratings
- reverse correlation (spike-triggered average)

Why do simple cells respond to bars and gratings of given orientation?



#### References to origins – modeling

#### 1D:

S. Marcelja: Mathematical description of the responses of simple cortical cells. Journal of the Optical Society of America 70, 1980, pp. 1297-1300.

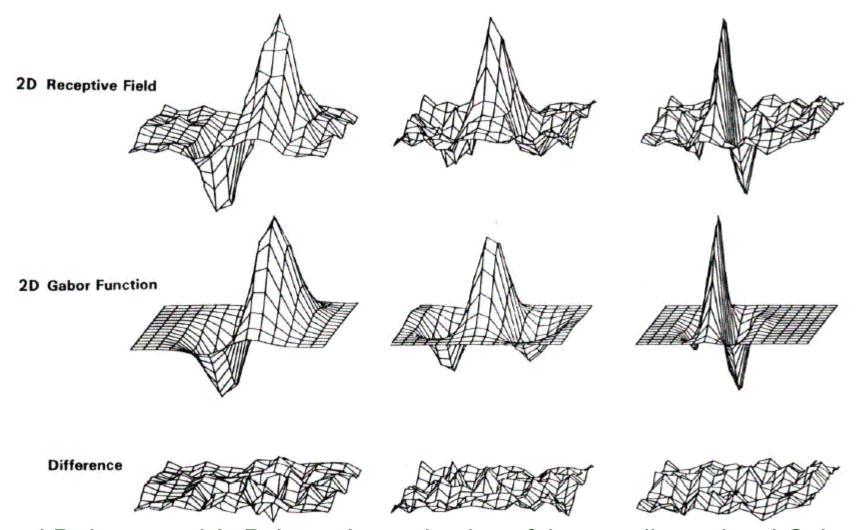
#### 2D:

J.G. Daugman: Uncertainty relations for resolution in space, spatial frequency, and orientation optimized by two-dimensional visual cortical filters, Journal of the Optical Society of America A, 1985, vol. 2, pp. 1160-1169.

J.P. Jones and A. Palmer: An evaluation of the two-dimensional Gabor filter model of simple receptive fields in cat striate cortex, Journal of Neurophysiology, vol. 58, no. 6, pp. 1233--1258, 1987



#### References to origins – modeling

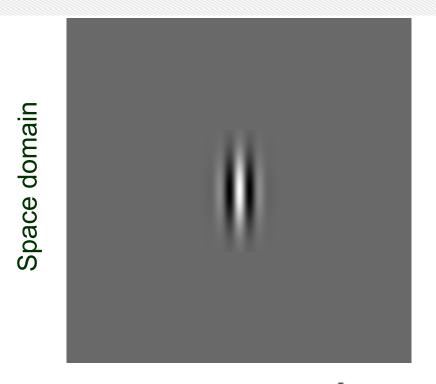


J.P. Jones and A. Palmer: An evaluation of the two-dimensional Gabor filter model of simple receptive fields in cat striate cortex, Journal of Neurophysiology, vol. 58, no. 6, pp. 1233--1258, 1987

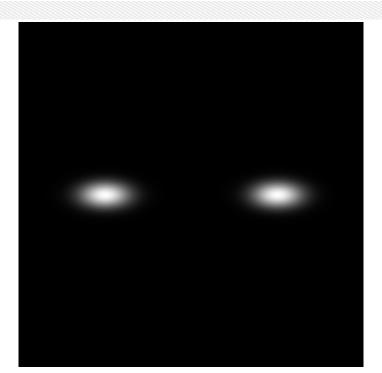


#### **2D Gabor functions**

#### 2D Gabor function



Frequency domain



$$g_{\lambda,\Theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-rac{x'^2 + \gamma^2 y'^2}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

$$egin{array}{ll} x' &=& x \mathrm{cos} \Theta + y \mathrm{sin} \Theta \ y' &=& -x \mathrm{sin} \Theta + y \mathrm{cos} \Theta \end{array}$$



#### Parameterization according to:

- N. Petkov: Biologically motivated computationally intensive approaches to image pattern recognition, *Future Generation Computer Systems*, **11** (4-5), 1995, 451-465.
- N. Petkov and P. Kruizinga: Computational models of visual neurons specialised in the detection of periodic and aperiodic oriented visual stimuli: bar and grating cells, *Biological Cybernetics*, **76** (2), 1997, 83-96.
- P. Kruizinga and N. Petkov: Non-linear operator for oriented texture, *IEEE Trans. on Image Processing*, **8** (10), 1999, 1395-1407.
- S.E. Grigorescu, N. Petkov and P. Kruizinga: Comparison of texture features based on Gabor filters, *IEEE Trans. on Image Processing*, **11** (10), 2002, 1160-1167.
- N. Petkov and M. A. Westenberg: Suppression of contour perception by band-limited noise and its relation to non-classical receptive field inhibition, *Biological Cybernetics*, **88**, 2003, 236-246.
- C. Grigorescu, N. Petkov and M. A. Westenberg: Contour detection based on nonclassical receptive field inhibition, *IEEE Trans. on Image Processing*, **12** (7), 2003, 729-739. http://www.cs.rug.nl/~petkov/publications/journals

#### Preferred spatial frequency $(1/\lambda)$ and size $(\sigma)$

$$g_{\lambda,\Theta,arphi, au,oldsymbol{\gamma}}(x,y) = \exp\left(-rac{x'^2 + \gamma^2 y'^2}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

$$x' = x\cos\Theta + y\sin\Theta$$
  
 $y' = -x\sin\Theta + y\cos\Theta$ 

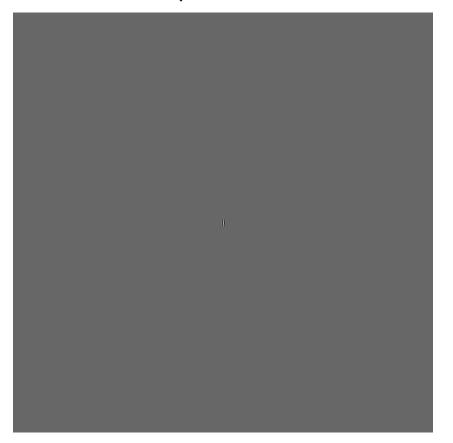
Preferred spatial frequency  $(1/\lambda)$  and size  $(\sigma)$  are not completely independent:

$$\sigma = a\lambda$$

with a between 0.3 and 0.6 for most cells. In the following, we use mostly  $\sigma = 0.56\lambda$ .



Space domain



Wavelength = 2/512

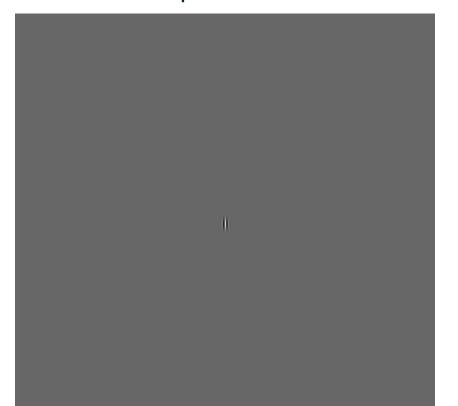
Frequency domain



Frequency = 512/2



Space domain



Wavelength = 4/512

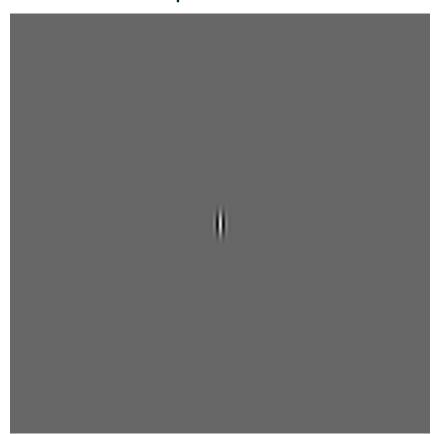
Frequency domain



Frequency = 512/4

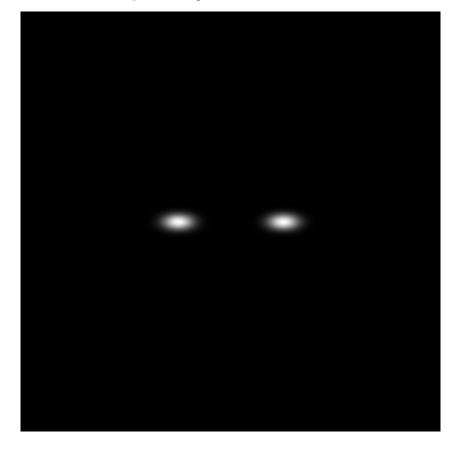


Space domain



Wavelength = 8/512

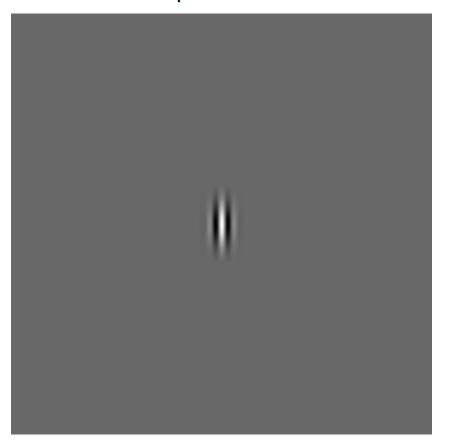
Frequency domain



Frequency = 512/8

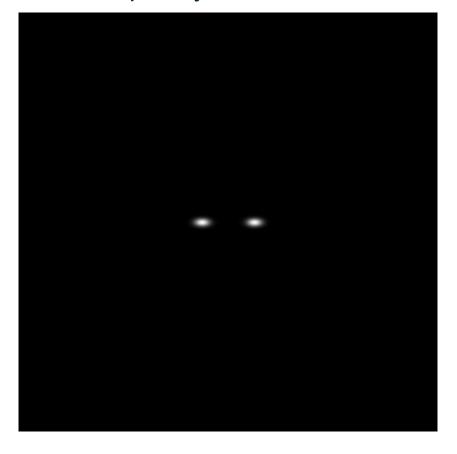


Space domain



Wavelength = 16/512

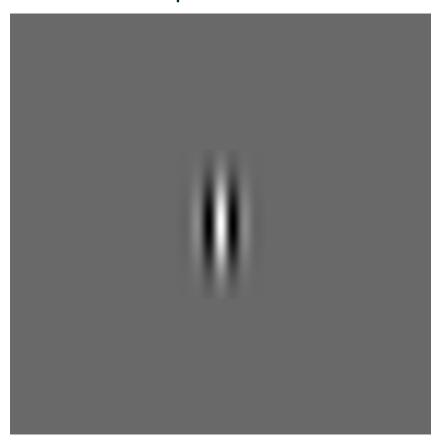
Frequency domain



Frequency = 512/16

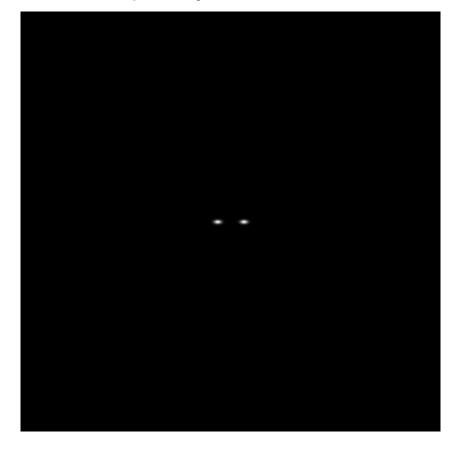


Space domain



Wavelength = 32/512

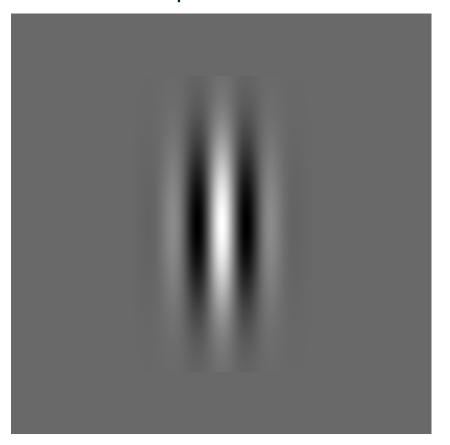
Frequency domain



Frequency = 512/32

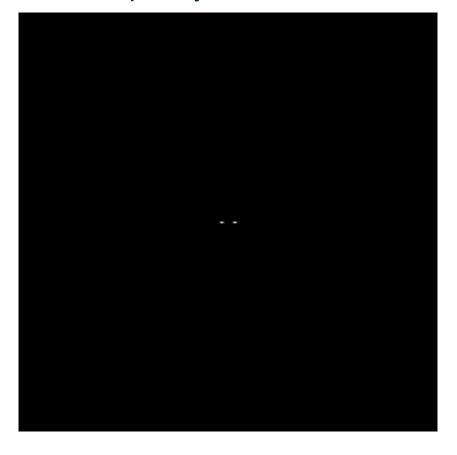


Space domain



Wavelength = 64/512

Frequency domain



Frequency = 512/64



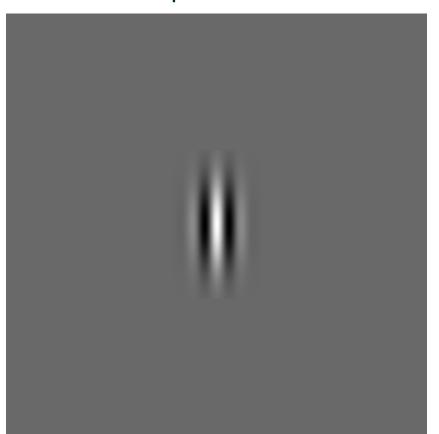


$$g_{\lambda,\Theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-rac{x'^2 + \gamma^2 y'^2}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

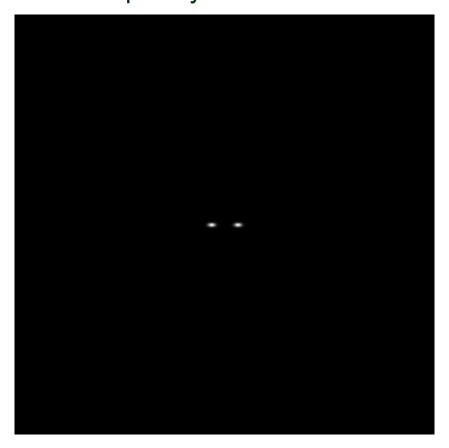
$$egin{array}{lll} x' &=& x \mathrm{cos} \Theta + y \mathrm{sin} \Theta \ y' &=& -x \mathrm{sin} \Theta + y \mathrm{cos} \Theta \end{array}$$



Space domain



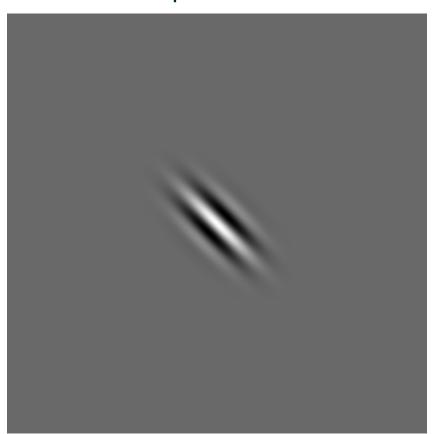
Frequency domain



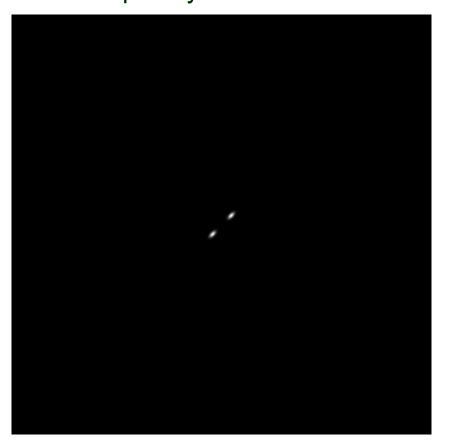
Orientation = 0



Space domain



Frequency domain

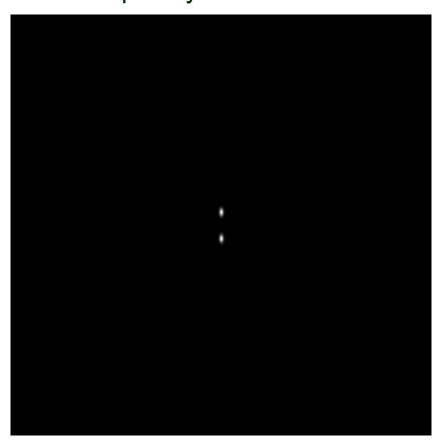


Orientation = 45



Space domain

Frequency domain



Orientation = 90

#### Symmetry (phase offset φ)

$$g_{\lambda,\Theta,arphi,\sigma,m{\gamma}}(x,y) = \exp\left(-rac{x''' + \gamma^2 y'''}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

$$egin{array}{lll} x' &=& x \mathrm{cos} \Theta + y \mathrm{sin} \Theta \ y' &=& -x \mathrm{sin} \Theta + y \mathrm{cos} \Theta \end{array}$$



#### Symmetry (phase offset)

Space domain



Phase offset = 0 (symmetric function)

Space domain



Phase offset = -90 (anti-symmetric function)

#### Spatial aspect ratio (γ)

$$g_{\lambda,\Theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-rac{x'^2 + \gamma^2 y'^2}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

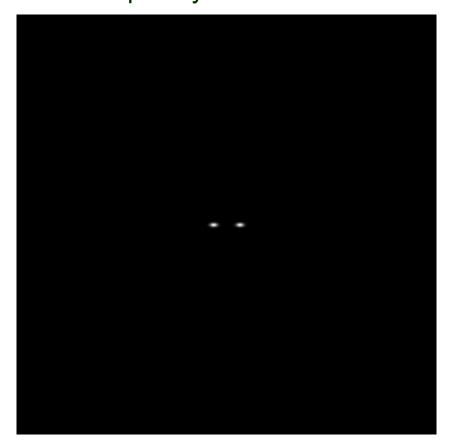
$$egin{array}{lll} x' &=& x \mathrm{cos} \Theta + y \mathrm{sin} \Theta \ y' &=& -x \mathrm{sin} \Theta + y \mathrm{cos} \Theta \end{array}$$



#### **Spatial aspect ratio**

Space domain

Frequency domain



Aspect ratio = 0.5



#### **Spatial aspect ratio**

Space domain

Frequency domain



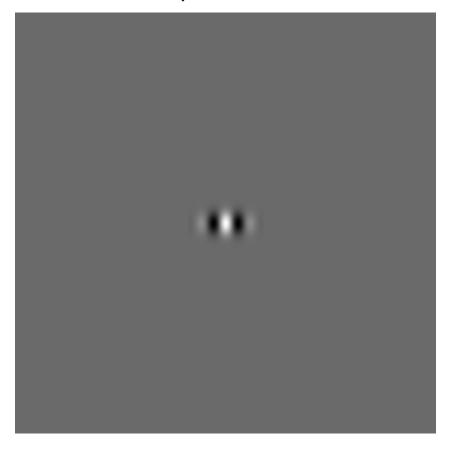
Aspect ratio = 1



#### **Spatial aspect ratio**

Space domain

Frequency domain





Aspect ratio = 2 (does not occur)

#### Bandwidth – related to the ratio $\sigma/\lambda$

#### Half-response spatial frequency bandwidth b (in octaves)

$$b = \log_2 \frac{\frac{\sigma}{\lambda}\pi + \sqrt{\frac{\ln 2}{2}}}{\frac{\sigma}{\lambda}\pi - \sqrt{\frac{\ln 2}{2}}}, \quad \frac{\sigma}{\lambda} = \frac{1}{\pi}\sqrt{\frac{\ln 2}{2}} \cdot \frac{2^b + 1}{2^b - 1} \quad (2)$$

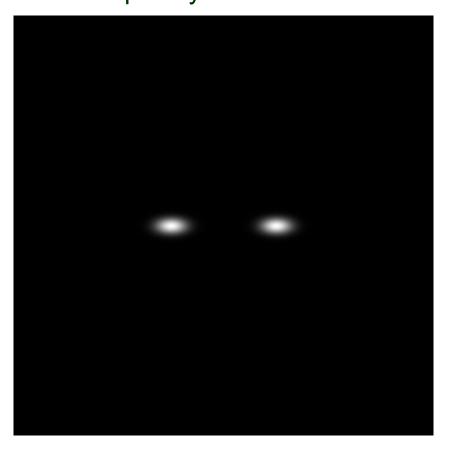
$$g_{\lambda,\Theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-rac{x'^2 + \gamma^2 y'^2}{2\sigma^2}
ight)\cos\left(2\pirac{x'}{\lambda} + arphi
ight) \quad (1)$$

$$egin{array}{lll} x' &=& x \cos \Theta + y \sin \Theta \ y' &=& -x \sin \Theta + y \cos \Theta \end{array}$$



Space domain

Frequency domain



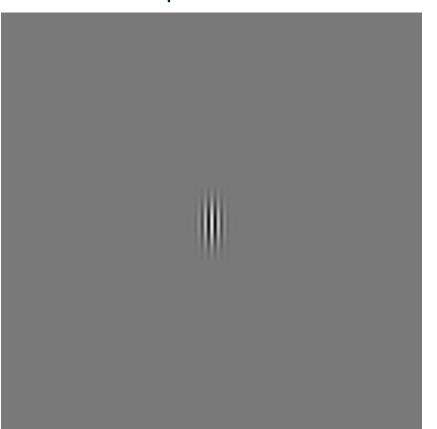
Bandwidth = 1 (
$$\sigma = 0.56\lambda$$
)

Wavelength = 8/512





Space domain



Frequency domain



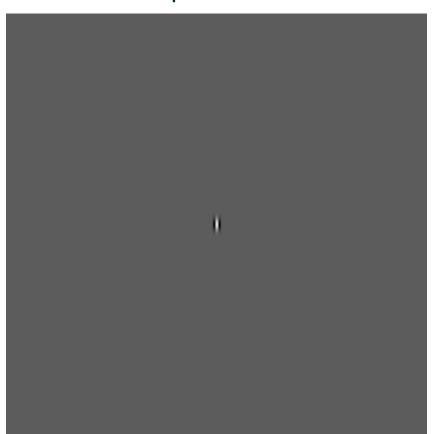
Bandwidth = 0.5

Wavelength = 8/512

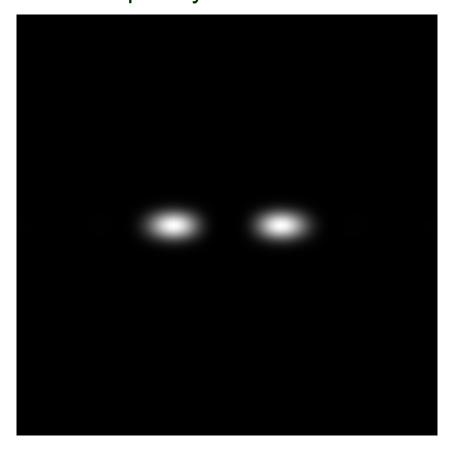




Space domain



Frequency domain

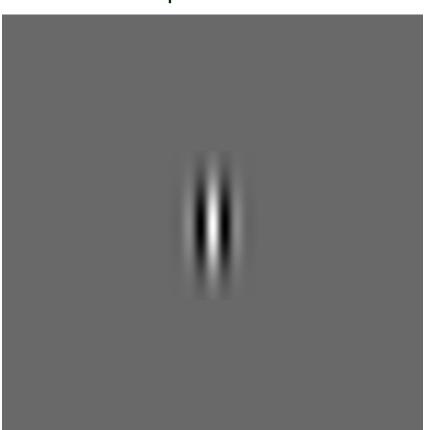


Bandwidth = 2

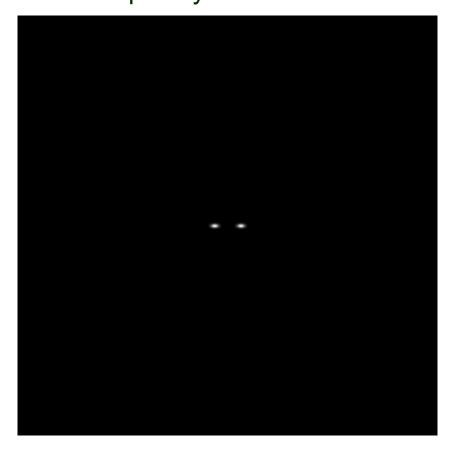
Wavelength = 8/512



Space domain



Frequency domain

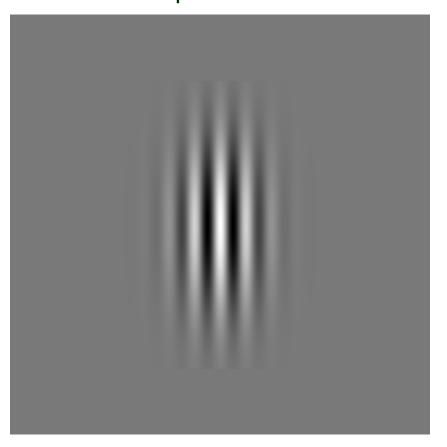


Bandwidth = 1 (
$$\sigma = 0.56\lambda$$
)  
Wavelength = 32/512





Space domain



Frequency domain



Bandwidth = 0.5

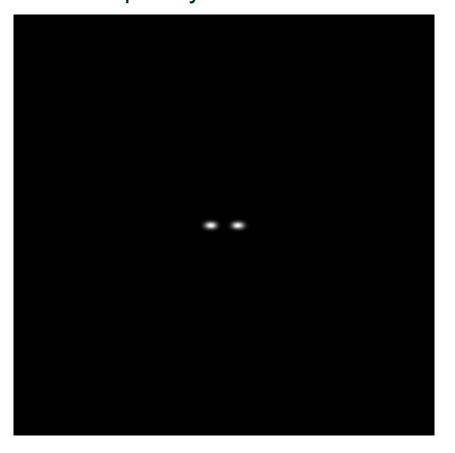
Wavelength = 32/512





Space domain

Frequency domain



Bandwidth = 2 Wavelength = 32/512



### Semi-linear 2D Gabor filter

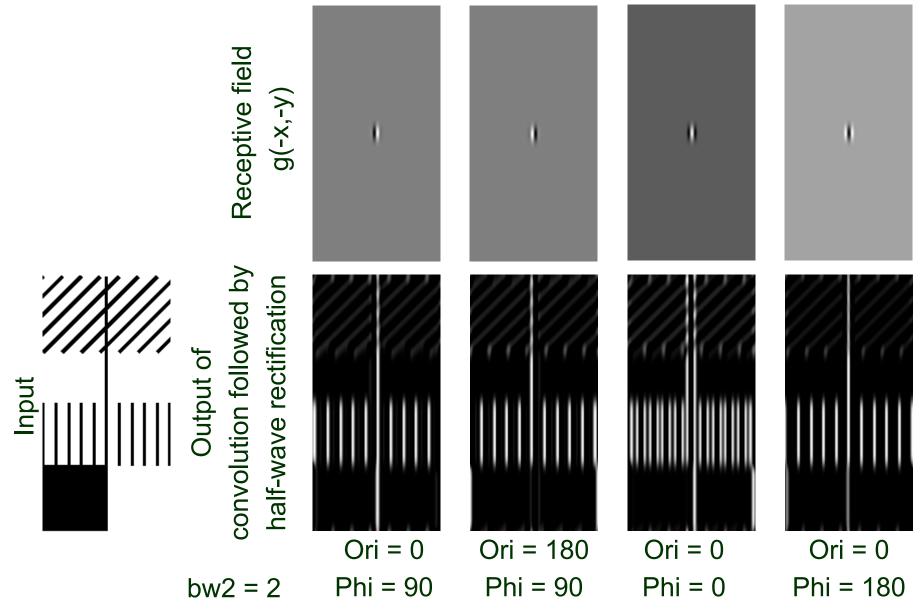
$$R = |g * I|^+$$

i.e., the response R is obtained by convolution (\*) of the input I with a Gabor function g, followed by half-wave rectification (|.|+)



### Semi-linear Gabor filter

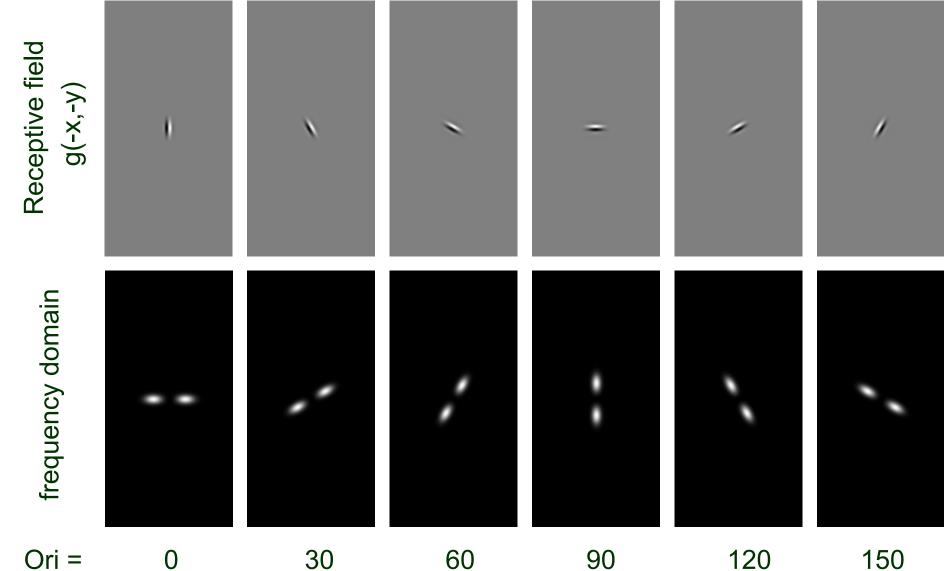
#### What is it useful for?



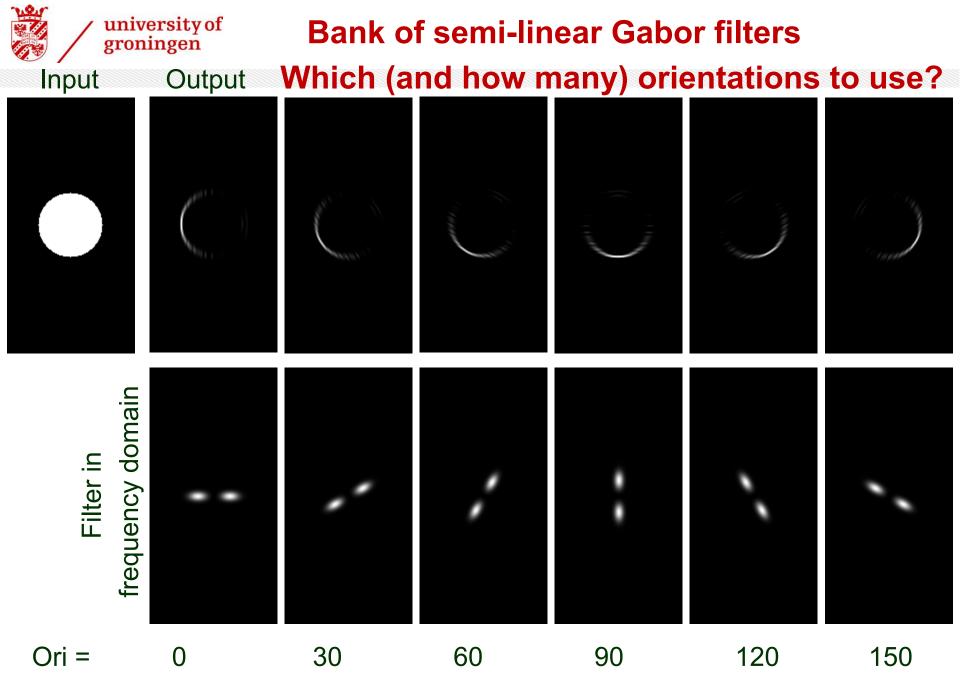
edaes lines lines



Which (an how many) orientations to use?



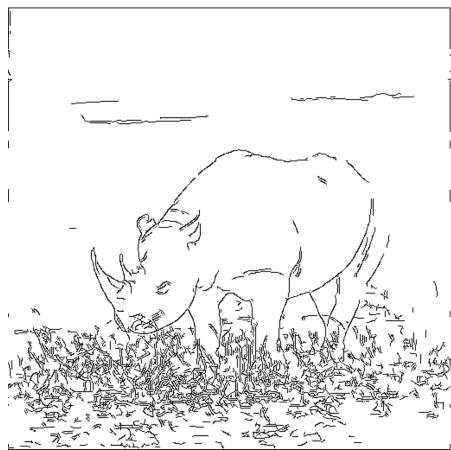
For filters with s.a.r=0.5 and bw=2, good coverage of angles with 6 orientations



For filters with sar=0.5 and bw=2, good coverage of angles with 12 orientations



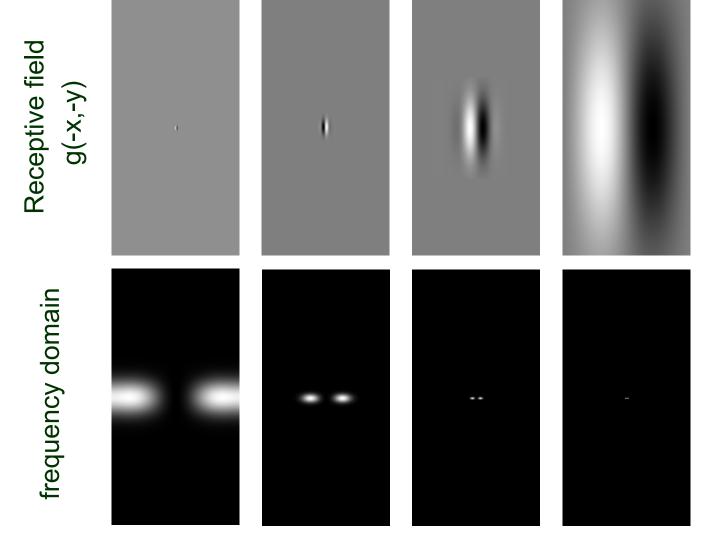




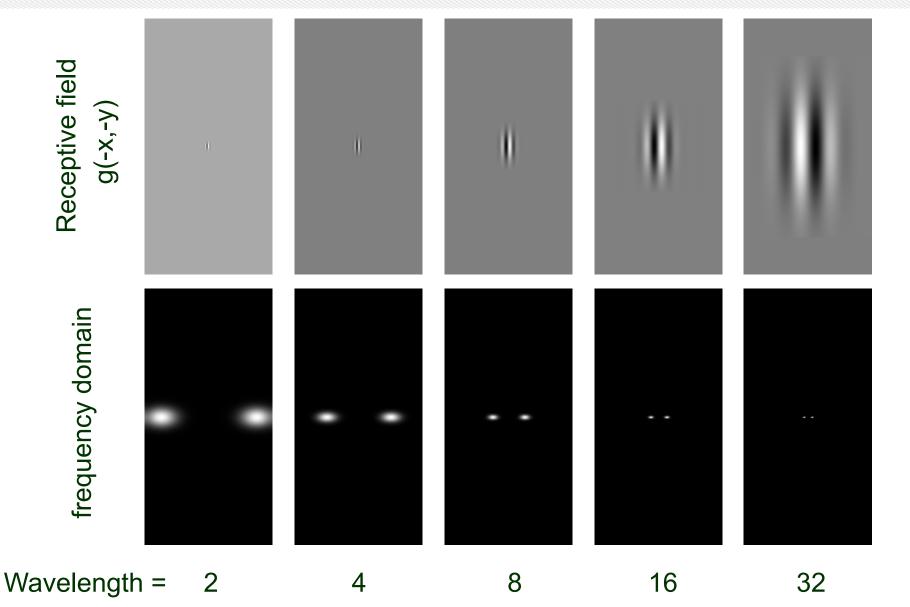
Result of superposition of the outputs of 12 semi-linear anti-symmetric (phi=90) Gabor filters with wavelength = 4, bandwidth = 2, spatial aspect ratio = 0.5 (after thinning and thresholding It = 0.1, ht = 0.15).



Which (and how many) frequencies to use?



Wavelength = 2 8 32 128 (s.a.r.=0.5) For filters with bw=2, good coverage of frequencies with wavelength quadroppling



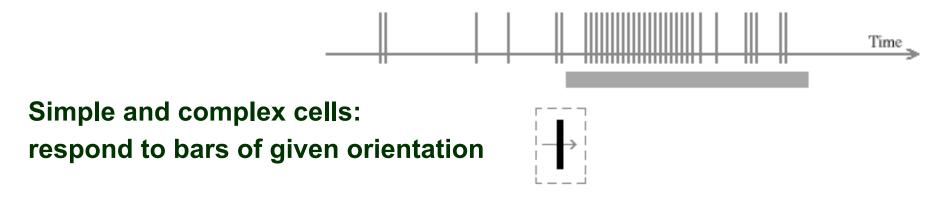
For filters with bw=1, good coverage of frequencies with wavelength doubling



# Complex cells and Gabor energy filters



#### References to origin - neurophysiology



D.H. Hubel and T.N. Wiesel: Receptive fields, binocular interaction and functional architecture in the cat's visual cortex, Journal of Physiology (London), vol. 160, pp. 106--154, 1962.

D.H. Hubel and T.N. Wiesel: Sequence regularity and geometry of orientation columns in the monkey striate cortex, Journal of Computational Neurology, vol. 158, pp. 267--293, 1974.

D.H. Hubel: Exploration of the primary visual cortex, 1955-78, Nature, vol. 299, pp. 515--524, 1982.

Hubel and Wiesel named another type of cell "complex" because they contrasted simple cells in the following properties:

- Their receptive fields do not have distinct excitatory and inhibitory regions.
- Their response cannot be predicted by weighted summation.
- Response is not modulated by the exact position of the optimal stimulus (bar or grating).

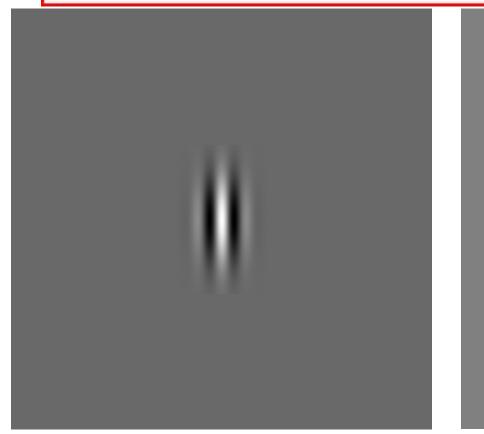
*In engineering terms*:

a complex cell cannot be characterized by an impulse response.



#### Gabor energy model of a complex cells

$$E_{\lambda,\sigma,\theta}(x,y) = \sqrt{R_{\lambda,\sigma,\theta,0}^2(x,y) + R_{\lambda,\sigma,\theta,-\frac{\pi}{2}}^2(x,y)}$$



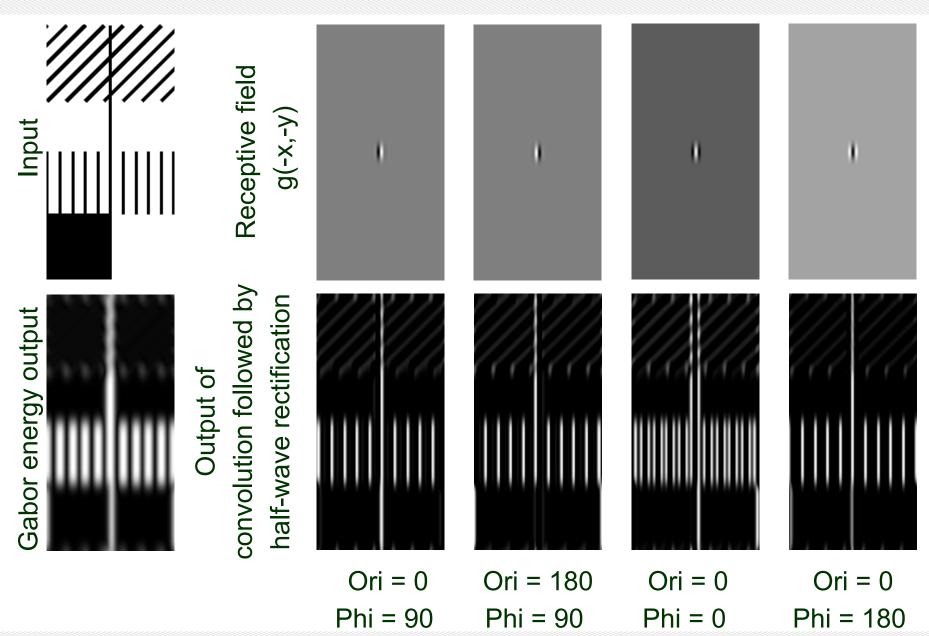
Phase offset = 0 (symmetric function)

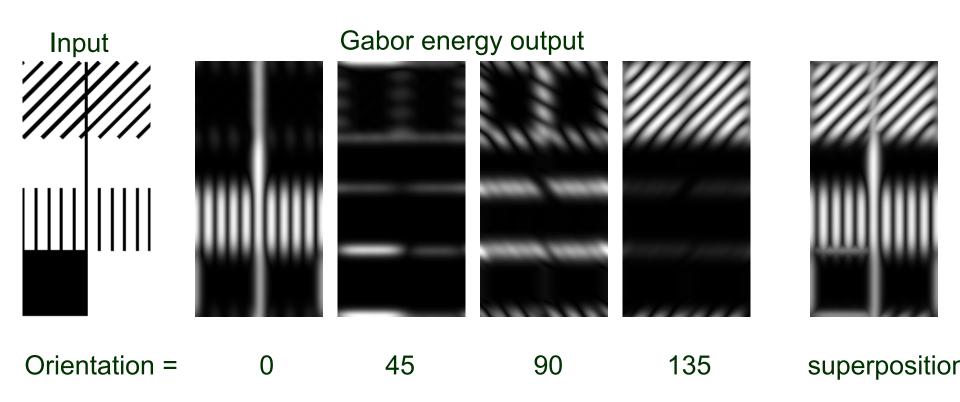


Phase offset = -90 (anti-symmetric function)



#### Gabor energy filter



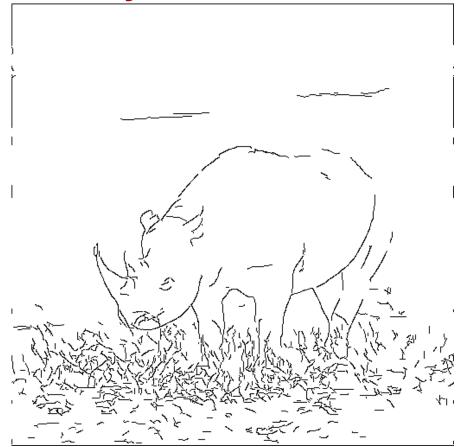


Result of superposition of the outputs of 4 Gabor energy filters (in [0,180)) with wavelength = 8, bandwidth = 1, spatial aspect ratio = 0.5



#### Bank of Gabor energy filters How many orientations to use?

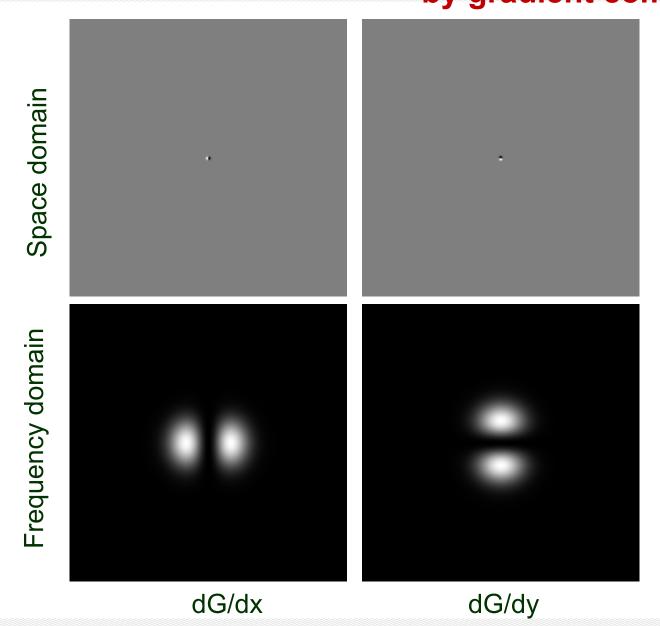




Result of superposition of the outputs of 6 Gabor energy filters (in [0,180)) with wavelength = 4, bandwidth = 2, spatial aspect ratio = 0.5 (after thinning and thresholding It = 0.1, ht = 0.15).



# More efficient way to detect intensity changes by gradient computation

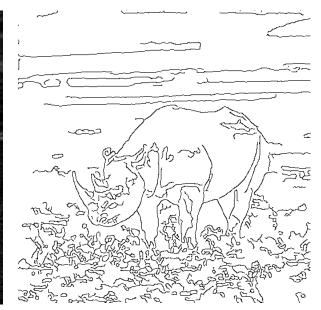




#### More efficient way to detect intensity changes





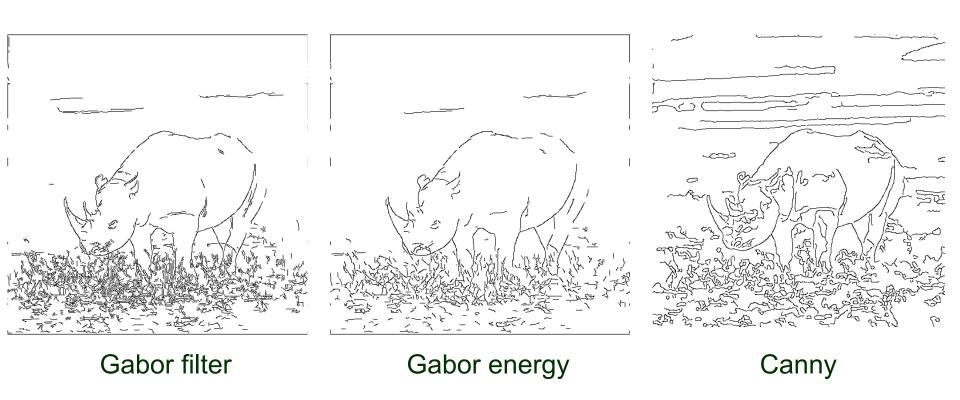


Gradient magnitude

Canny



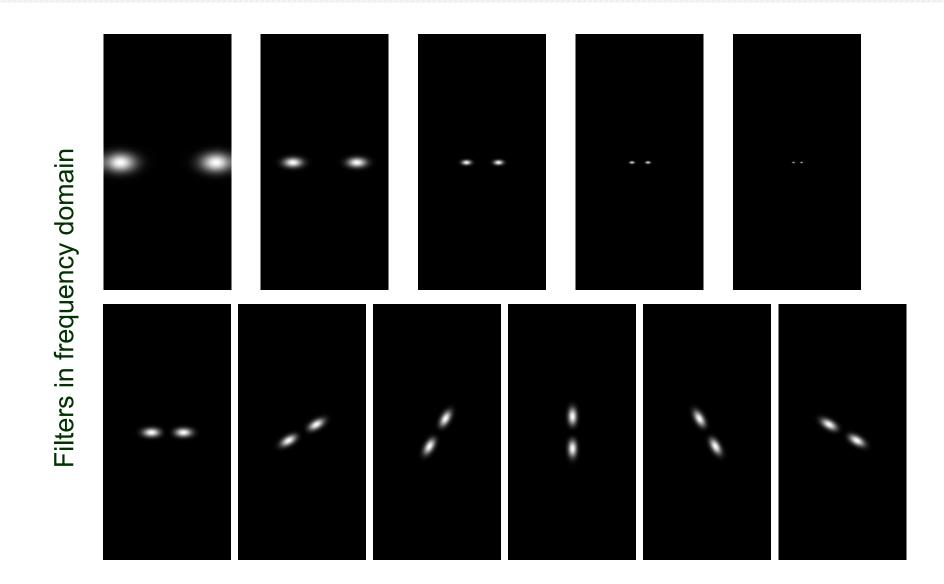
#### Various ways to detect edges



http://matlabserver.cs.rug.nl



#### **Gabor filters for texture analysis**





#### Gabor filters for texture analysis

See e.g.

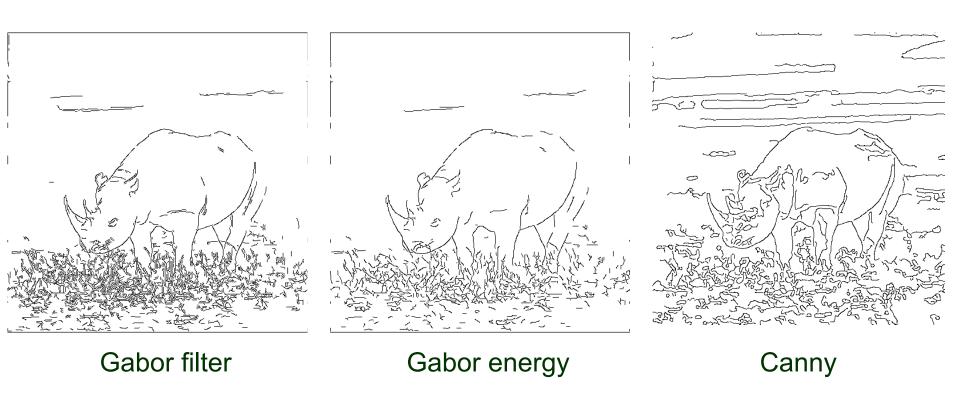
S.E. Grigorescu, N. Petkov and P. Kruizinga: Comparison of texture features based on Gabor filters, *IEEE Trans. on Image Processing*, **11** (10), 2002, 1160-1167.

and references therein

http://matlabserver.cs.rug.nl



#### **Problems with texture edges**

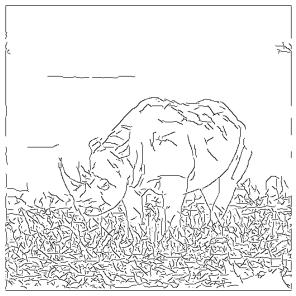


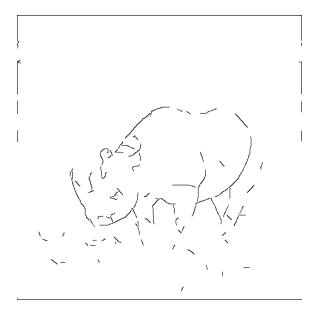
http://matlabserver.cs.rug.nl



#### Contour enhancement by suppression of texture







Canny

with surround suppression

[Petkov and Westenberg, Biol.Cyb. 2003] [Grigorescu et al., IEEE-TIP 2003, IVC 2004]



#### Spatiotemporal (3D) Gabor filters

#### See

N. Petkov and E. Subramanian:

Motion detection, noise reduction, texture suppression and contour enhancement by spatiotemporal Gabor filters with surround inhibition,

Biological Cybernetics, 97 (5-6), 2007, 423-439.

and references therein

http://www.cs.rug.nl/~petkov/publications/journals

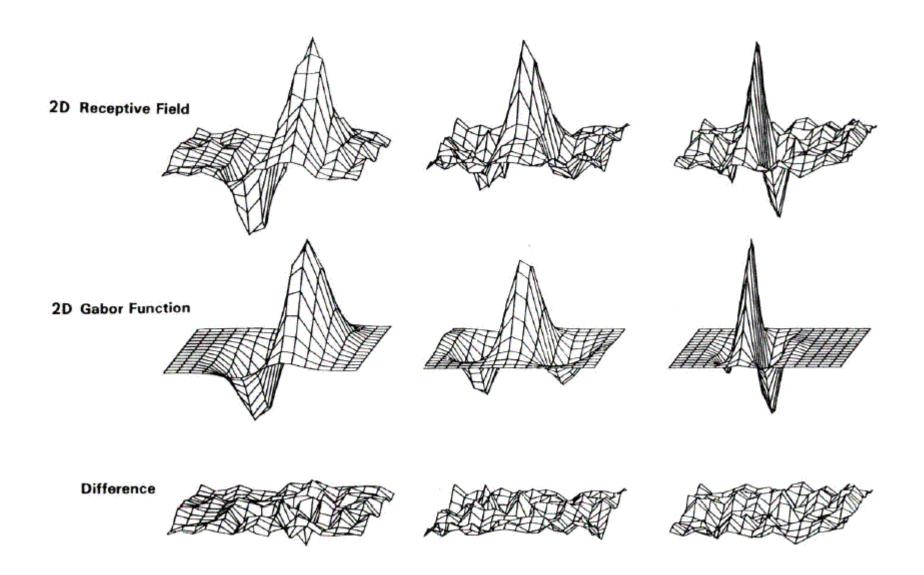


# Complex cells and CORF filters

(or a non-Gaborian, less Platonic view of reality)

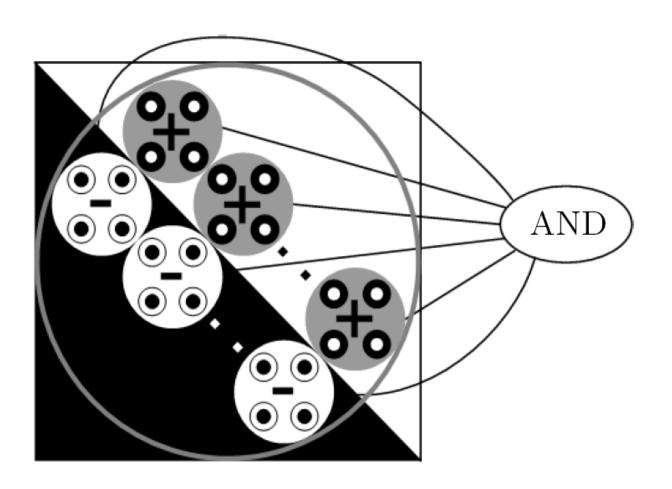


#### References to origins – modeling





**CORF: Combination Of LGN Receptive Fields** 



[Azzopardi and Petkov, 2011]



**CORF: Combination Of LGN Receptive Fields** 

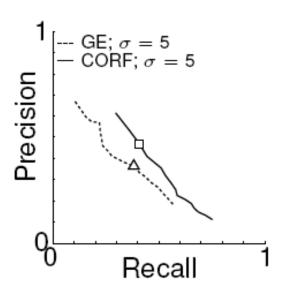




Image

Ground truth





Gabor Energy

CORF



**CORF: Combination Of LGN Receptive Fields** 



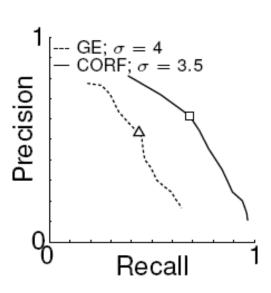


Image

Ground truth



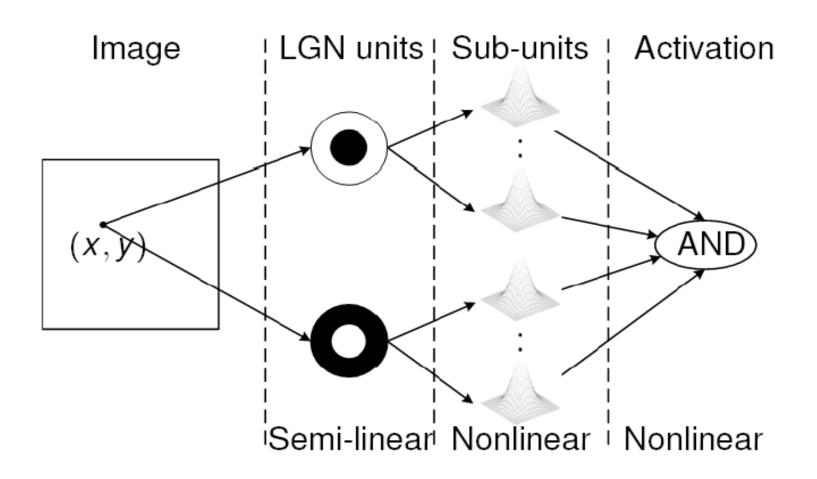




Gabor Energy

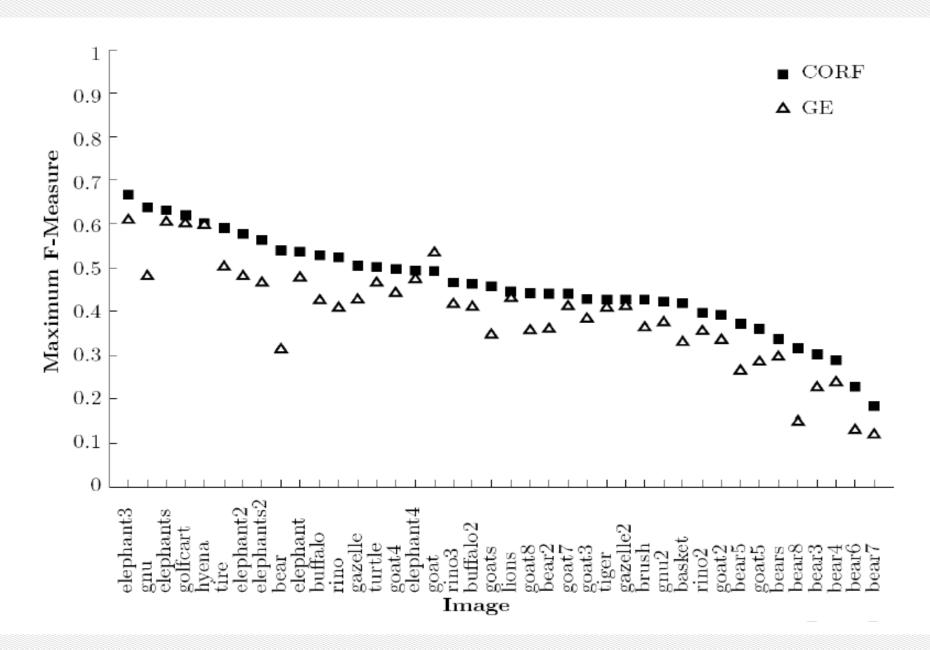
CORF

**CORF: Combination Of LGN Receptive Fields** 











# V1 complex cells modeled by CORF: Combination Of LGN Receptive Fields

# (a) $SNR = \infty$ (b) SNR = 5(c) SNR = 2.5(e) CORF (d) CORF (f) CORF F-Measure = 0.72 F-Measure = 0.51 F-Measure = 1 (g) GE (h) GE (i) GE F-Measure = 0.48 F-Measure = 0.43 F-Measure = 0.17

#### CORF is more effective than GE

- Better contour integration
- More robust to noise
- Better edge localization





Computational models (of V1/2) are approximations

Plato: use Gabor energy (for aesthetic reasons)

Popper: use CORF (for practical empiricism)