# Pattern Classification (EET 3035)

Extra Lecture: 2D-Gabor Filters

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#### Outline

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#### Introduction

Introduction 0000

- Gabor filters are bandpass filters which are used in image processing for
  - feature extraction.
  - texture analysis,
  - stereo disparity estimation, etc.
- The kernel mask of these filters is created by multiplying a Gaussian envelop function with a complex oscillation.

$$g_{\lambda,\theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-\frac{x'^2 + \gamma^2 y'^2}{2\sigma^2}\right) \exp\left(i\left(2\pi\frac{x'}{\lambda} + \varphi\right)\right)$$

It was shown by several researchers that the profile of simple-cell receptive fields in the mammalian visual cortex can by described by oriented two-dimensional Gabor functions.

#### Visual Cortex

Introduction 0000

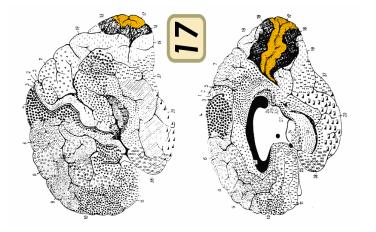
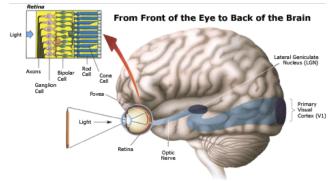


Figure: Broadmann area 171

#### Visual System

Introduction 0000

- In the spatial domain, a 2D Gabor filter is a Gaussian kernel function modulated by a sinusoidal plane wave.
- Some authors claim that simple cells in the visual cortex of mammalian brains can be modelled by Gabor functions.
- Simple cells respond to bars and gratings of given orientation.



#### 2D Gabor Functions

Complex

$$g_{\lambda,\theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-\frac{x'^2 + \gamma^2 y'^2}{2\sigma^2}\right) \exp\left(i\left(2\pi\frac{x'}{\lambda} + \varphi\right)\right)$$

Real

Introduction

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$$g_{\lambda,\theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-\frac{x'^2 + \gamma^2 y'^2}{2\sigma^2}\right) \cos\left(2\pi \frac{x'}{\lambda} + \varphi\right)$$

**Imaginary** 

$$g_{\lambda,\theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-\frac{x'^2 + \gamma^2 y'^2}{2\sigma^2}\right) \sin\left(2\pi \frac{x'}{\lambda} + \varphi\right)$$

where

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

### 2D Gabor function parameters

$$g_{\lambda,\theta,\varphi,\sigma,\gamma}(x,y) = \exp\left(-\frac{x'^2 + \gamma^2 y'^2}{2\sigma^2}\right) \cos\left(2\pi \frac{x'}{\lambda} + \varphi\right)$$

where

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

 $\lambda \rightarrow$  wavelength of the sinusoidal factor,

 $\theta \rightarrow$  orientation of the normal to the parallel stripes of a Gabor function.

 $\varphi \to \mathsf{phase}$  offset,

 $\sigma \rightarrow$  standard deviation of the Gaussian envelope,

 $\gamma \to \text{spatial}$  aspect ratio, specifies the ellipticity of the support of the Gabor function.

## Wavelength $(\lambda)$

- Wavelength of the cosine factor of the Gabor filter kernel.
- Value is to be specified in number of pixels.
- Valid values are real numbers,  $\lambda \ge 2$
- The value  $\lambda=2$  should not be used in combination with phase offset  $\phi=-90$  or  $\phi=90$ , because in these cases the Gabor function is sampled in its zero crossings.
- To avoid undesirable effect at the image borders,  $\lambda$  value should be smaller than one fifth of the input size.

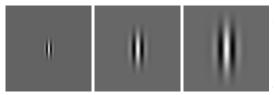


Figure: Image size is  $100\times100,~\lambda=5,10,15$  from left to right, other parameters  $\theta=0,~\varphi=0,~\gamma=0.5,~b=1$ 

## Orientations $(\theta)$

- This parameter specifies the orientation of the normal to the parallel stripes of a Gabor function.
- Value is to be specified in degrees.
- Valid values are real numbers between 0-360.

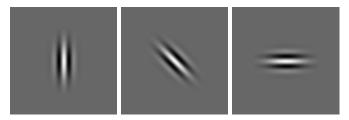


Figure: Image size is  $100 \times 100$ ,  $\theta = 0, 45, 90$  from left to right, other parameters  $\lambda = 10, \ \varphi = 0, \ \gamma = 0.5, \ b = 1$ 

## Phase offset $(\varphi)$

- The phase offset  $\varphi$  in the cosine factor of the Gabor function is specified in degree.
- Valid values are real number in between -180 and 180.
- The values 0 and 180 corresponds to center-symmetric 'center-on' and 'center-off' function, respectively. While -90 and 90corresponds to anti-symmetric functions

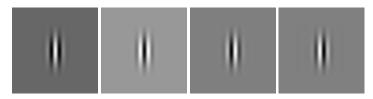


Figure: Image size is  $100 \times 100$ ,  $\phi = 0$ , 180, -90, and 90 degrees from left to right, other parameters  $\lambda = 10$ ,  $\theta = 0$ ,  $\gamma = 0.5$ , b = 1

## Aspect ratio $(\gamma)$

- This is spatial aspect ratio which specifies the ellipticity of the support of the Gabor function.
- For  $\gamma = 1$ , the support is circular.
- For  $\gamma < 1$  the support is elongated in orientation of the parallel stripes of the function
- Normal value is  $\gamma = 0.5$

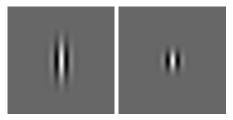


Figure: Image size is  $100 \times 100$ ,  $\gamma = 0.5$  and 1 degrees from left to right, other parameters  $\lambda = 10$ ,  $\theta = 0$ ,  $\varphi = 0$ , b = 1

## Bandwidth (b)

■ The half-response spatial frequency bandwidth b (in octaves) of a Gabor filter is related to the ratio  $\frac{\sigma}{\lambda}$ .

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

- $\blacksquare$  The value of  $\sigma$  cannot be specified directly. It can only be changed through the bandwidth b.
- Must be a real positive number.
- The smaller the bandwidth, the larger  $\sigma$ ,

# Spatial frequency $(1/\lambda)$

• Preferred spatial frequency,  $1/\lambda$ , and size  $\sigma$  are not fully independent. Values are related with a relation

$$\sigma = a\lambda$$

- a varies in between 0.03 and 0.6 for most cells.
- In many experiments, a=0.56 is used, i.e.,  $\sigma=0.56\lambda$ .

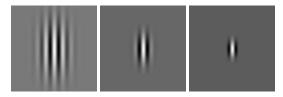
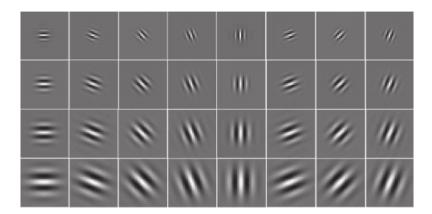
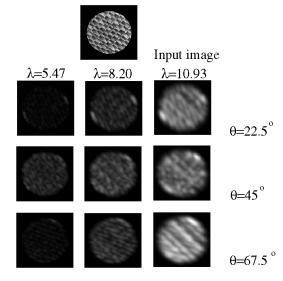


Figure: Image size is  $100 \times 100$ , b = 0.5, 1 and 2 from left to right, respectively. Other parameters  $\lambda=10,~\theta=0,~\varphi=0,~\gamma=0.5$ 

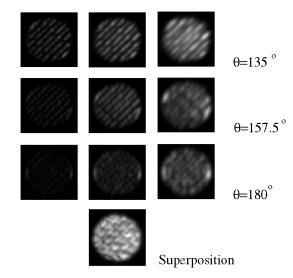


Applications •000

# Application - Texture Segmentation



Applications 0000



# Application - Texture Segmentation

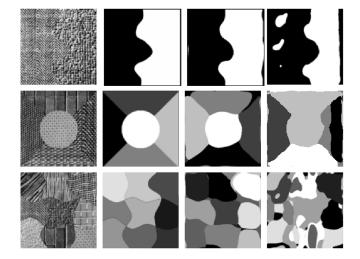


Figure: Results of segmentation experiments using the K-means clustering algorithm.

#### References

- [1] John G Daugman. "Uncertainty relation for resolution in space, spatial frequency, and orientation optimized by two-dimensional visual cortical filters". In: *JOSA A* 2.7 (1985), pp. 1160–1169.
- [2] Cosmin Grigorescu, Nicolai Petkov, and Michel A Westenberg. "Contour detection based on nonclassical receptive field inhibition". In: *IEEE Transactions on image processing* 12.7 (2003), pp. 729–739.
- [3] Simona E Grigorescu, Nicolai Petkov, and Peter Kruizinga. "Comparison of texture features based on Gabor filters". In: *IEEE Transactions on Image processing* 11.10 (2002), pp. 1160–1167.

#### References

- [4] Anil K Jain and Farshid Farrokhnia. "Unsupervised texture segmentation using Gabor filters". In: Pattern recognition 24.12 (1991), pp. 1167–1186.
- Peter Kruizinga, Nicolai Petkov, and Simona E Grigorescu. [5] "Comparison of texture features based on Gabor filters". In: Image analysis and processing, 1999. Proceedings. International conference on. IEEE. 1999, pp. 142-147.
- [6] Nicolai Petkov and Michel A Westenberg. "Suppression of contour perception by band-limited noise and its relation to nonclassical receptive field inhibition". In: Biological cybernetics 88.3 (2003), pp. 236–246.

#### References

- [7] Nikolai Petkov and Peter Kruizinga. "Computational models of visual neurons specialised in the detection of periodic and aperiodic oriented visual stimuli: bar and grating cells". In: Biological cybernetics 76.2 (1997), pp. 83–96.
- [8] Nikolay Petkov. "Biologically motivated computationally intensive approaches to image pattern recognition". In: Future Generation Computer Systems 11.4-5 (1995), pp. 451-465.
- [9] Visual Cortex. Wikipedia, the free encyclopedia, https://en.wikipedia.org/wiki/Visual\_cortex.

