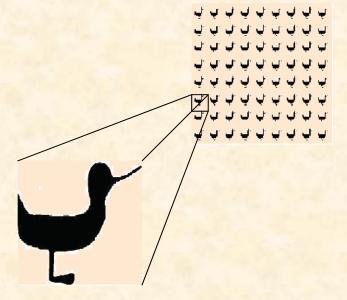
TEXTURE ANALYSIS

USING

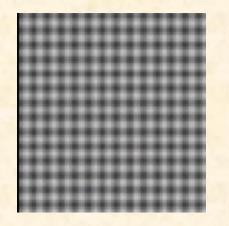
GABOR FILTERS

Texture Types

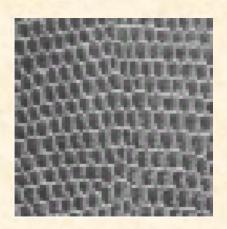
Definition of Texture



Texture types



Synthetic



Natural



Stochastic

Texture Definition

Texture: the regular repetition of an element or pattern on a surface.



- Purpose of texture analysis:
 - To identify different textured and nontextured regions in an image.
 - To classify/segment different texture regions in an image.
 - To extract boundaries between major texture regions.
 - To describe the texel unit.
 - 3-D shape from texture

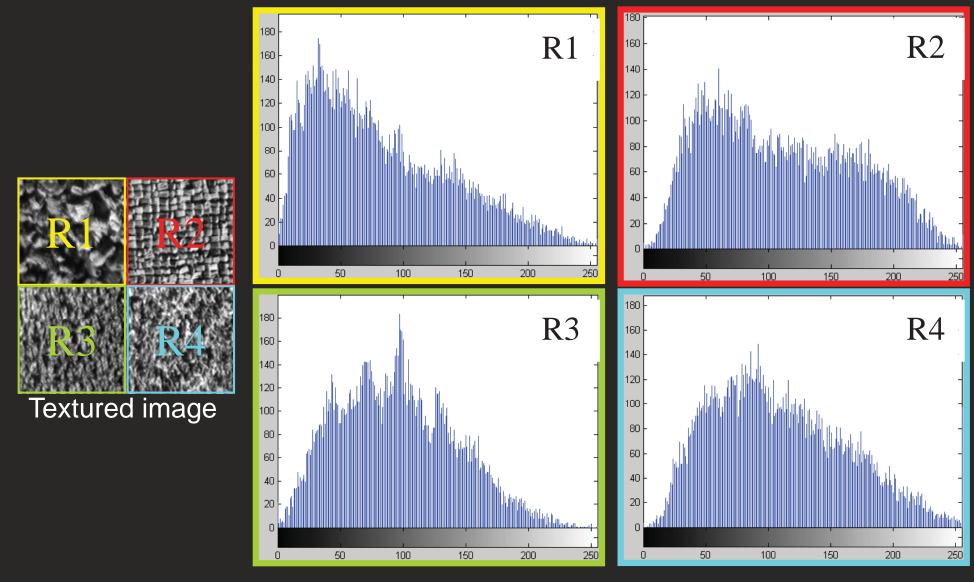
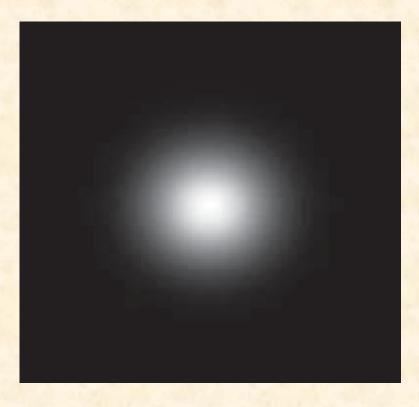


Image histograms

Processing of Texture-like Images

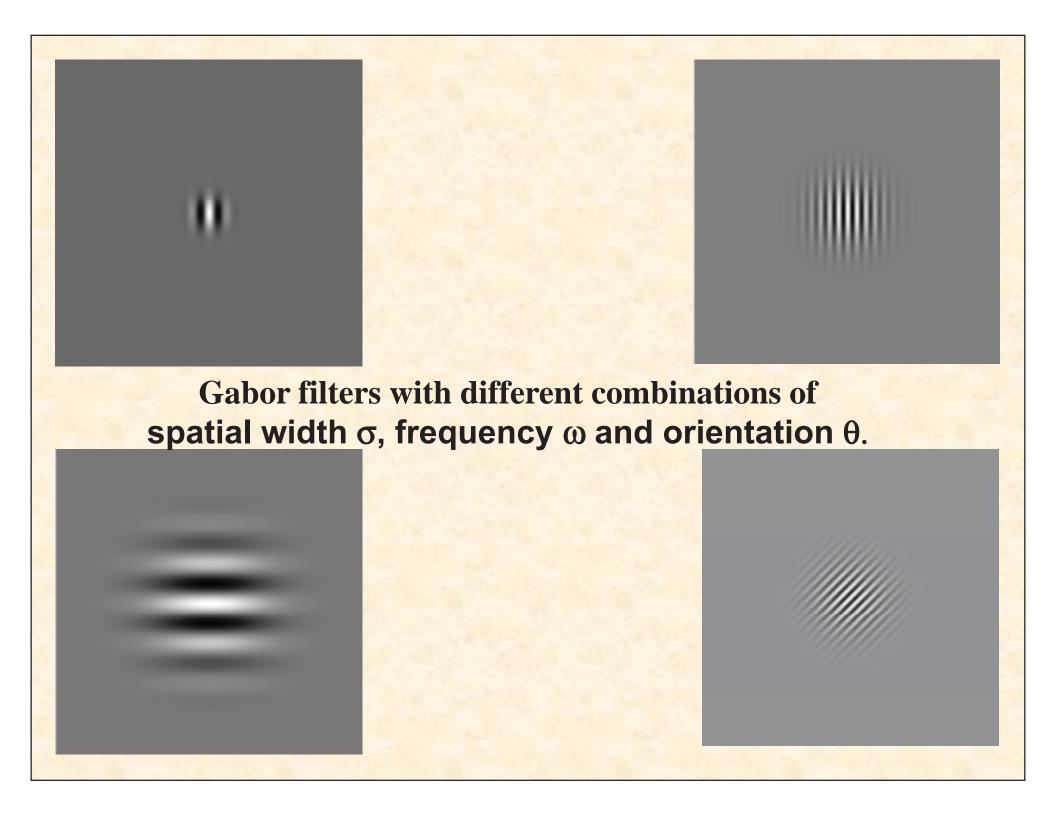
2-D Gabor Filter

$$f(x, y, \omega, \theta, \sigma_x, \sigma_y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp\left[\frac{-1}{2}\left(\left(\frac{x}{\sigma_x}\right)^2 + \left(\frac{y}{\sigma_y}\right)^2\right) + j\omega(x\cos\theta + y\sin\theta)\right]$$



A typical Gaussian filter with $\sigma=30$

A typical Gabor filter with σ =30, ω =3.14 and θ =45°



2-D Gabor filter:

$$f(x, y, \omega, \theta, \sigma_x, \sigma_y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp\left[\frac{-1}{2}((\frac{x}{\sigma_x})^2 + (\frac{y}{\sigma_y})^2) + j\omega(x\cos\theta + y\sin\theta)\right]$$

where

σ is the spatial spread

ω is the frequency

 θ is the orientation

1-D Gabor filter:

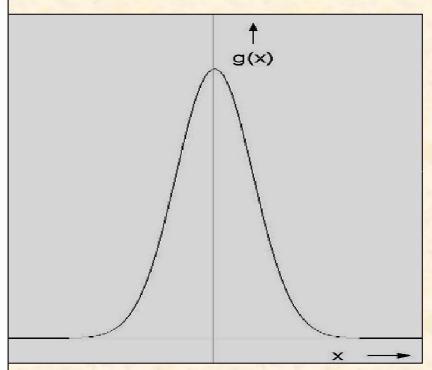
$$f(x,\omega,\sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp(\frac{-x^2}{2\sigma^2} + j\omega x)$$

$$g(x) = \frac{1}{\sqrt{2\pi\sigma_1}} \exp\left(\frac{-x^2}{2\sigma^2}\right)$$

Processing of Texture-like Images

1-D Gaussian Filter

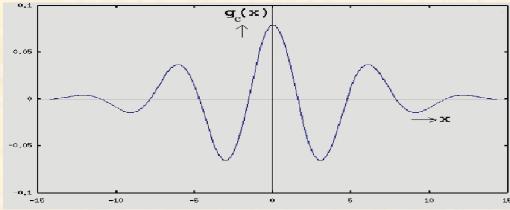
$$g(x) = \frac{1}{\sqrt{2\pi}\sigma_1} \exp\left(\frac{-x^2}{2\sigma^2}\right)$$



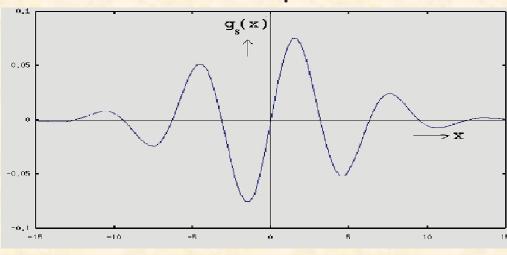
1-D Gabor Filter

$$f(x,\omega,\sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp(\frac{-x^2}{2\sigma^2} + j\omega x)$$

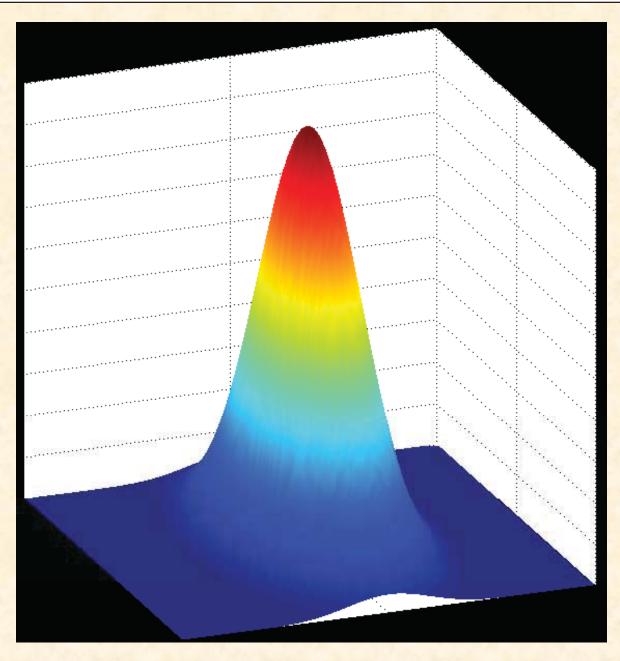
Even Component



Odd Component

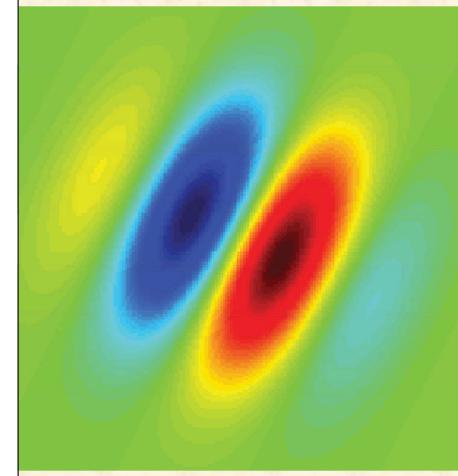






Asymmetric 2-D Gaussian function

$$gab(x, y) = K \exp(-\pi (a^2 (x - x_0)_{\theta}^2 + b^2 (y - y_0)_{\theta}^2))$$
$$\exp(j(2\pi F_0 (x \cos \omega_0 + y \sin \omega_0) + P)$$



gab(x, y) =

• **K**: Scales the magnitude of the Gaussian envelop.

• (a, b): Scale the two axis of the Gaussian envelop.

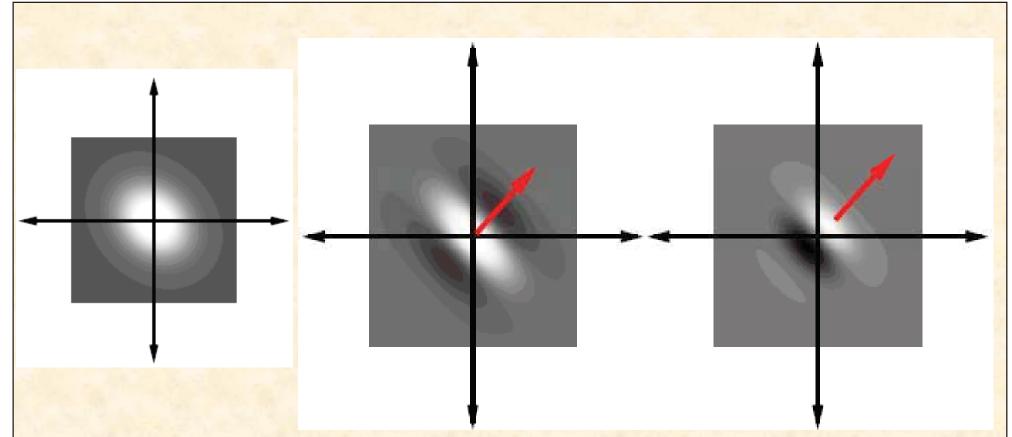
• θ : (Rotation) angle of the Gaussian envelop.

• $(\mathbf{x_0}, \mathbf{y_0})$: Location of the peak of the Gaussian envelop.

• $(\mathbf{u}_0, \mathbf{v}_0)$: Spatial frequencies of the sinusoidal carrier in Cartesian coordinates. It can also be expressed in polar coordinates as $(\mathbf{F}_0, \boldsymbol{\omega}_0)$.

• P: Phase of the sinusoidal carrier.

$$K \exp(-\pi (a^2(x-x_0)_{\theta}^2+b^2(y-y_0)_{\theta}^2)+j(2\pi(u_0x+v_0y)+P)$$

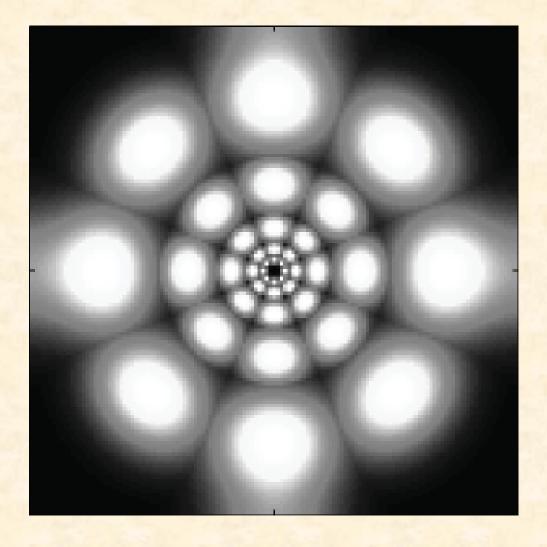


Asymmetrical Gaussian of 128×128 pixels. The parameters are as follows:

 $x_0 = y_0 = 0$; a = 1/50 pixels; b = 1/40 pixels; $\theta = -45$ deg.

The real and imaginary parts of a complex Gabor function in space domain, with

 $F_0 = \text{sqrt}(2)/80 \text{ cycles/pixel}, \ \omega_0 = 45 \text{ deg}, \ P = 0 \text{ deg}.$

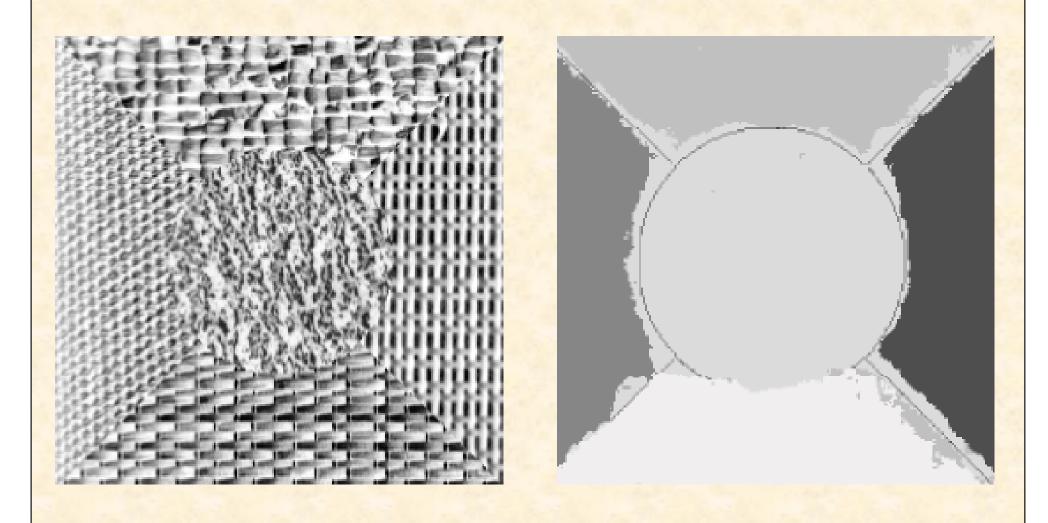


The frequency response of a typical <u>dyadic</u> bank of Gabor filters.

One center-symmetric pair of lobes in the illustration represents each filter.

Some properties of Gabor filters:

- A tunable bandpass filter
- Similar to a STFT or windowed Fourier transform
- Satisfies the lower-most bound of the time-spectrum resolution (uncertainty principle)
- It's a multi-scale, multi-resolution filter
- Has selectivity for orientation, spectral bandwidth and spatial extent.
- Has response similar to that of the Human visual cortex (first few layers of brain cells)
- Used in many applications texture segmentation; iris, face and fingerprint recognition.
- Computational cost often high, due to the necessity of using a large bank of filters (or Gabor jet) in most applications



Segmentation using Gabor based features of a texture image containing five regions.