

2D Fourier Transform for Images

Digital Signal and Image Processing

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Outline

2D Fourier Transform

Fourier Filtering

2D Fourier Transform

The 2D Fourier Transform can be easily derived from the 1D.

► **Fourier Transform:**

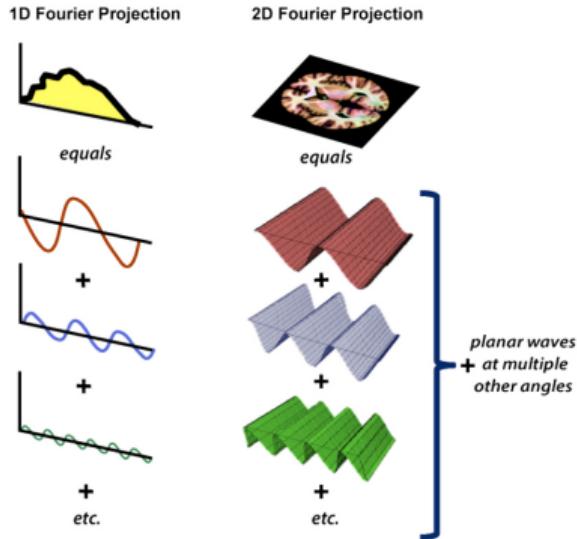
$$F(\omega_1, \omega_2) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) e^{-i(\omega_1 x + \omega_2 y)} dx dy$$

► **Inverse of the Fourier Transform:**

$$f(x, y) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} F(\omega_1, \omega_2) e^{i(\omega_1 x + \omega_2 y)} d\omega_1 d\omega_2$$

ω_1 and ω_2 are spatial frequencies.

Sinusoidal waves

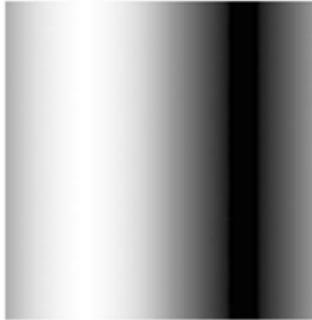


- ▶ Similarly to the 1D case, the FT is based on a decomposition into sinusoidal functions that form an orthonormal basis

$$e^{-i(\omega_1 x + \omega_2 y)} = \cos(\omega_1 x + \omega_2 y) + i \sin(\omega_1 x + \omega_2 y)$$

- ▶ the terms are sinusoids on the x, y plane

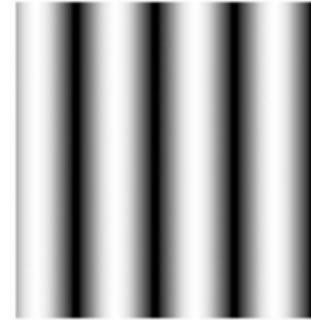
y
x



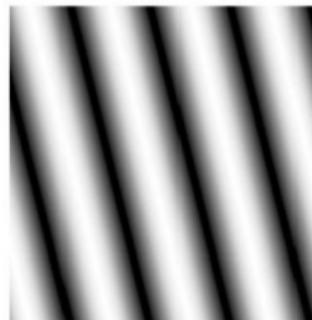
$$\omega_1 = 1, \omega_2 = 0$$



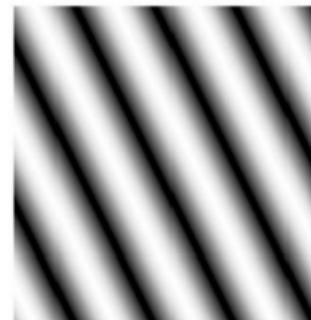
$$\omega_1 = 2, \omega_2 = 0$$



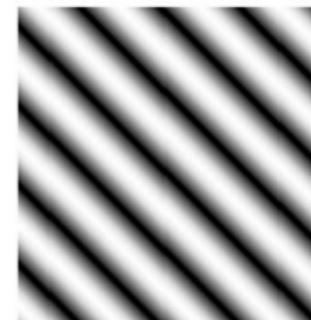
$$\omega_1 = 4, \omega_2 = 0$$



$$\omega_1 = 4, \omega_2 = 1$$



$$\omega_1 = 4, \omega_2 = 2$$



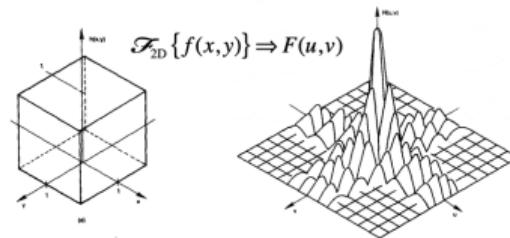
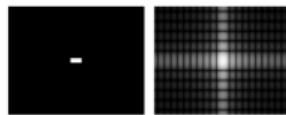
$$\omega_1 = 4, \omega_2 = 4$$

Putting things together...

$$f(x,y) = \alpha \begin{array}{c} \text{Image of a building} \\ \text{represented as a barcode} \end{array} + \beta \begin{array}{c} \text{Image of a building} \\ \text{represented as a barcode} \end{array} + \gamma \begin{array}{c} \text{Image of a building} \\ \text{represented as a diagonal hatched pattern} \end{array} + \dots$$

Example 1 - FT pairs

rectangle centred at origin
with sides of length X and Y



$$\begin{aligned}F(\omega_1, \omega_2) &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) e^{-i(\omega_1 x + \omega_2 y)} dx dy \\&= \int_{-X/2}^{X/2} e^{-i\omega_1 x} dx \int_{-Y/2}^{Y/2} e^{-i\omega_2 y} dy = \\&= \left[\frac{e^{-i\omega_1 x}}{-i\omega_1} \right]_{-X/2}^{X/2} \left[\frac{e^{-i\omega_2 y}}{-i\omega_2} \right]_{-Y/2}^{Y/2} = \\&= -\frac{1}{2i\omega_1} (e^{-i\omega_1 X} - e^{i\omega_1 X}) \left(-\frac{1}{2i\omega_2} \right) (e^{-i\omega_2 Y} - e^{i\omega_2 Y}) = \\&= \frac{1}{2} XY \frac{\sin(X\omega_1)}{X\omega_1} \frac{\sin(Y\omega_2)}{Y\omega_2} = \frac{1}{2} XY \text{sinc}(X\omega_1) \text{sinc}(Y\omega_2)\end{aligned}$$

FT pair example 2

Gaussian centred on origin

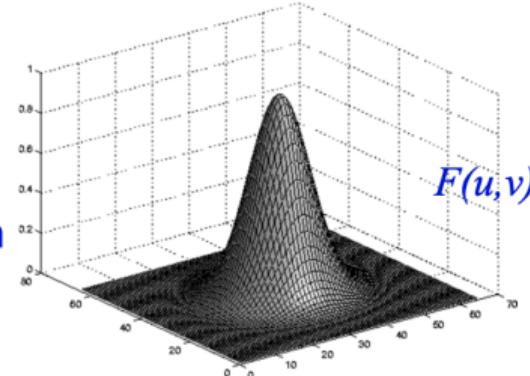
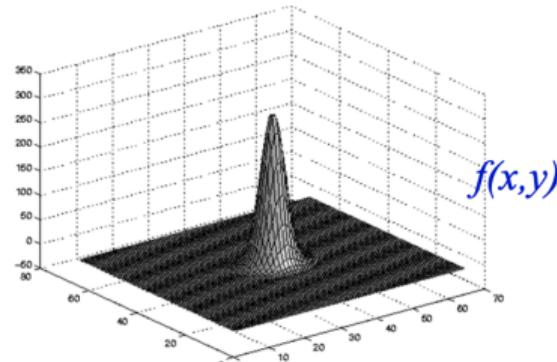
$$f(r) = \frac{1}{2\pi\sigma^2} e^{-r^2/2\sigma^2}$$

where $r^2 = x^2 + y^2$.

$$F(u, v) = F(\rho) = e^{-2\pi^2\rho^2\sigma^2}$$

where $\rho^2 = u^2 + v^2$.

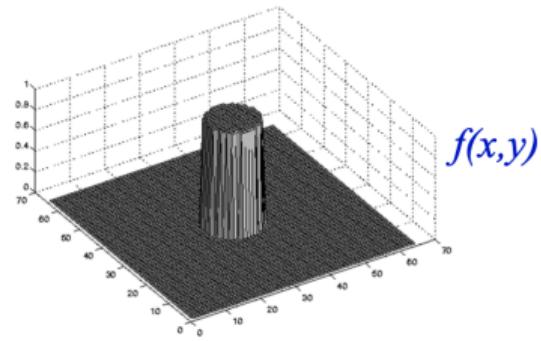
- FT of a Gaussian is a Gaussian
- Note inverse scale relation



FT pair example 3

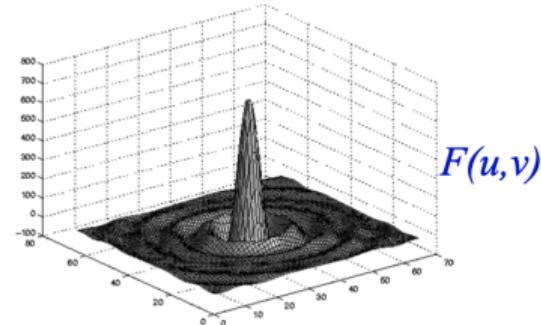
Circular disk unit height and radius a centred on origin

$$f(x, y) = \begin{cases} 1, & |r| < a, \\ 0, & |r| \geq a. \end{cases}$$



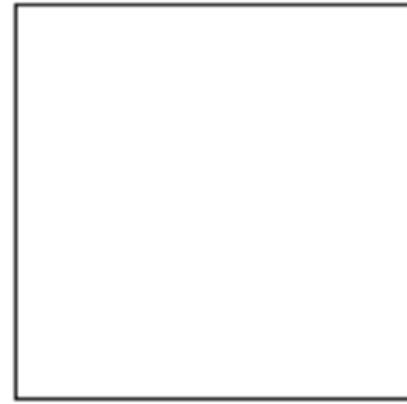
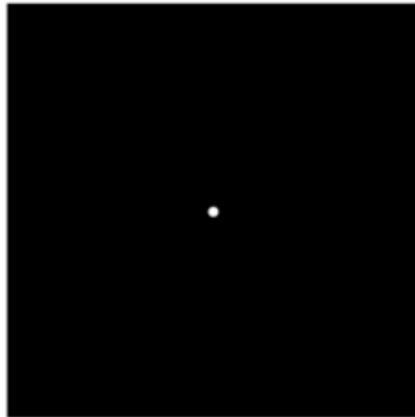
$$F(u, v) = F(\rho) = a J_1(\pi a \rho) / \rho$$

where $J_1(x)$ is a Bessel function.



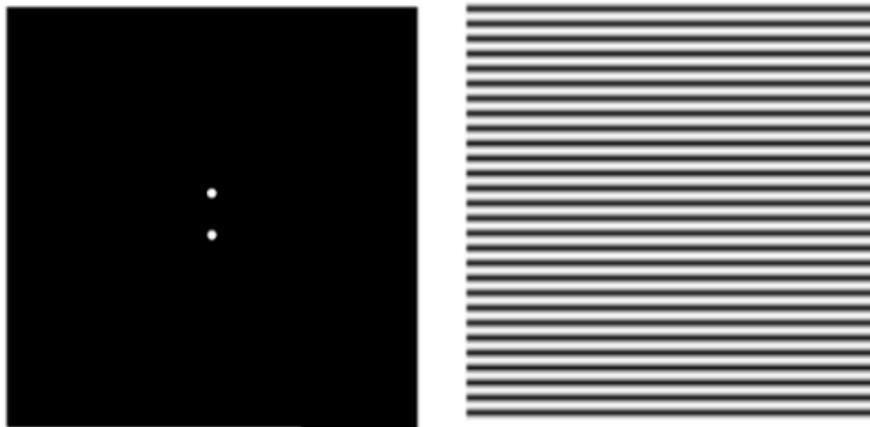
- rotational symmetry
- a '2D' version of a sinc

FT pairs



$$\begin{aligned}f(x,y) &= \delta(x,y) = \delta(x)\delta(y) \\F(\omega_1, \omega_2) &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \delta(x,y) e^{-i(\omega_1 x + \omega_2 y)} dx dy = 1\end{aligned}$$

FT pairs



$$\begin{aligned}f(x,y) &= \frac{1}{2}(\delta(x, y - a) + \delta(x, y + a)) \\F(\omega_1, \omega_2) &= \frac{1}{2} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \delta(x, y - a) + \delta(x, y + a) e^{-i(\omega_1 x + \omega_2 y)} dx dy = \\&= \frac{1}{2}(e^{-ia\omega_2} + e^{ia\omega_2}) = \cos(a\omega_2)\end{aligned}$$

Complex Conjugate symmetry of a real function

- ▶ An image is usually a 2D real-valued signal.
- ▶ If $f(x, y)$ is real, $f(x, y) = f^*(x, y)$, then $F^*(\omega_1, \omega_2) = F(-\omega_1, -\omega_2)$;
- ▶ It follows that $|F(\omega_1, \omega_2)| = |F(-\omega_1, -\omega_2)|$, which says that the spectrum of the Fourier transform is symmetric.
- ▶ In other words, there exist negative frequencies which are mirror images of the corresponding positive frequencies .

Properties: Rotation in space

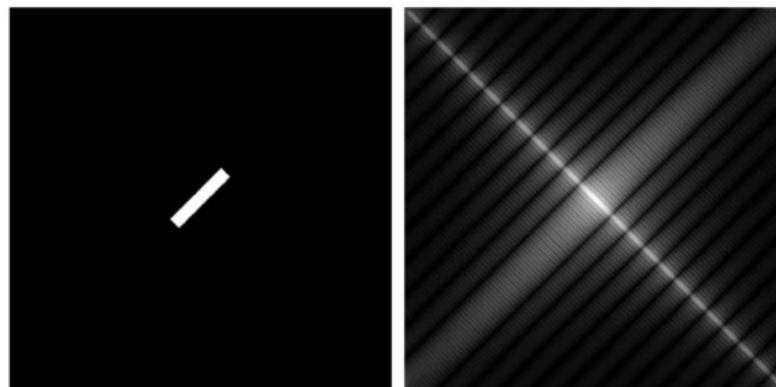
Let us assume g to be an in-plane rotation of f by an angle θ , that is $(x', y') = R_\theta(x, y)$

Then, $g(x, y) = f(x \cos \theta + y \sin \theta, -x \sin \theta + y \cos \theta)$

By the linearity property of the FT

$G(\omega_1, \omega_2) = F(\omega_1 \cos \theta - \omega_2 \sin \theta, \omega_1 \sin \theta + \omega_2 \cos \theta)$

The FT rotates by the same angle of a rotated image



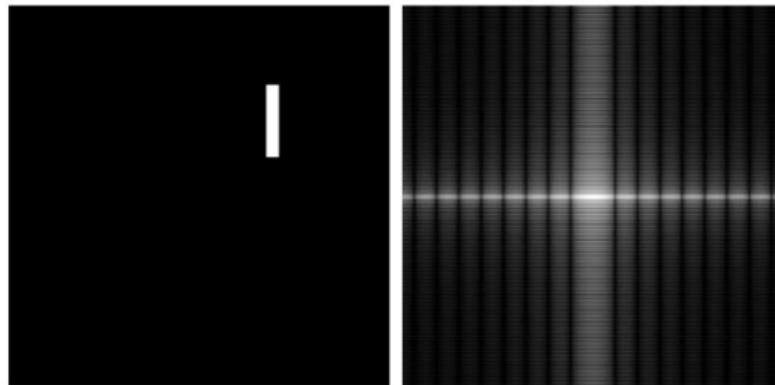
Properties: translation in space

(FT properties), Given $g(x - x_0, y - y_0)$, its FT $G(\omega_1, \omega_2)$ is related to $F(\omega_1, \omega_2)$ by a phase factor:

$$G(\omega_1, \omega_2) = e^{-i(x_0\omega_1 + y_0\omega_2)} F(\omega_1, \omega_2)$$

As for the spectrum: $|G(w)| = |e^{-i(x_0\omega_1 + y_0\omega_2)}| |F(\omega_1, \omega_2)| = |F(\omega_1, \omega_2)|$

The spectrum $|F(\omega_1, \omega_2)|$ is insensitive to image translation



Properties : shift in frequency

(Notice: duality)

given $G(\omega_1 + o_1, \omega_2 + o_2)$, then its IFT $g(x, y)$ is related to $f(x, y)$ by a phase factor

$$g(x, y) = e^{i(o_1 x + o_2 y)} f(x, y)$$

Outline

2D Fourier Transform

Fourier Filtering

Fourier Filtering

To filter an image in the frequency domain:

- ▶ Compute the DFT of the image $F(u, v)$
- ▶ Select and appropriate filter H (or filter transfer function)
- ▶ Multiply $F(u, v)$ by a filter function $H(u, v)$:

$$G(u, v) = H(u, v)F(u, v)$$

- ▶ Compute the inverse DFT of $G(u, v)$

Frequency domain filtering operation

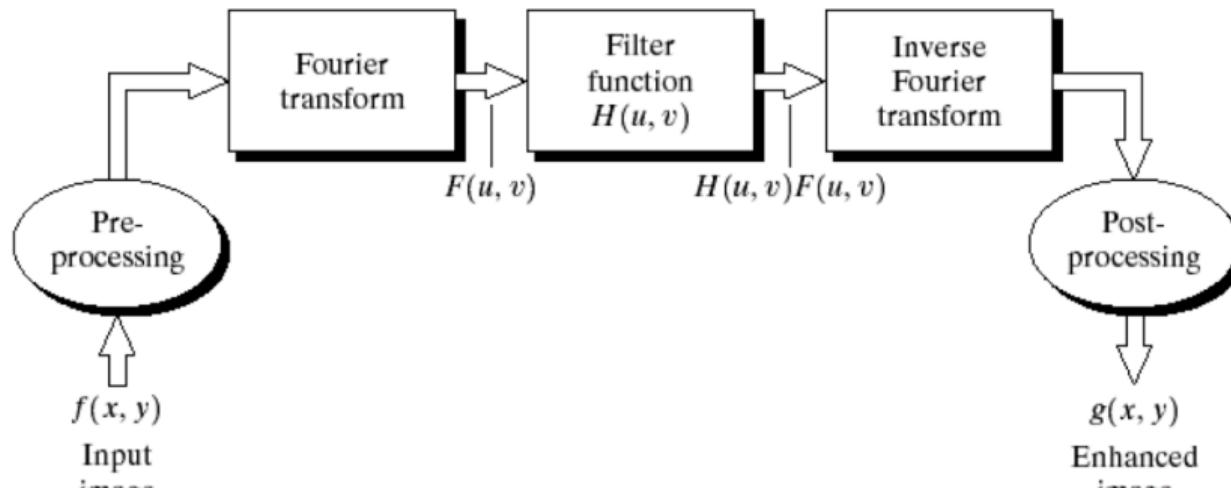
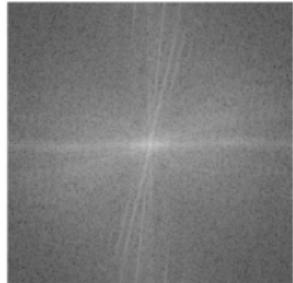


Image filtering and convolution theorem

$f(x,y)$



Fourier transform

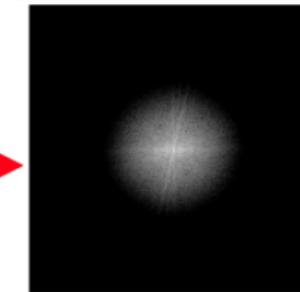


$|F(u,v)|$

$g(x,y)$



Inverse Fourier
transform



$|G(u,v)|$

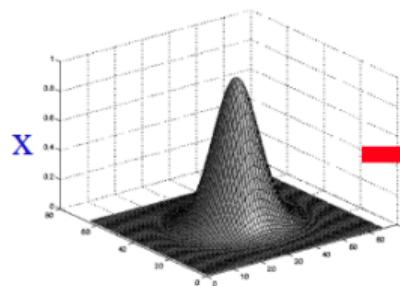


Image filtering and convolution theorem

$f(x,y)$



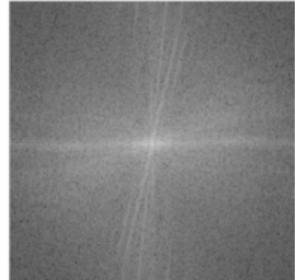
Gaussian
scale=3 pixels

*

$g(x,y)$

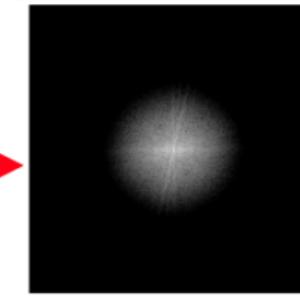


Fourier transform



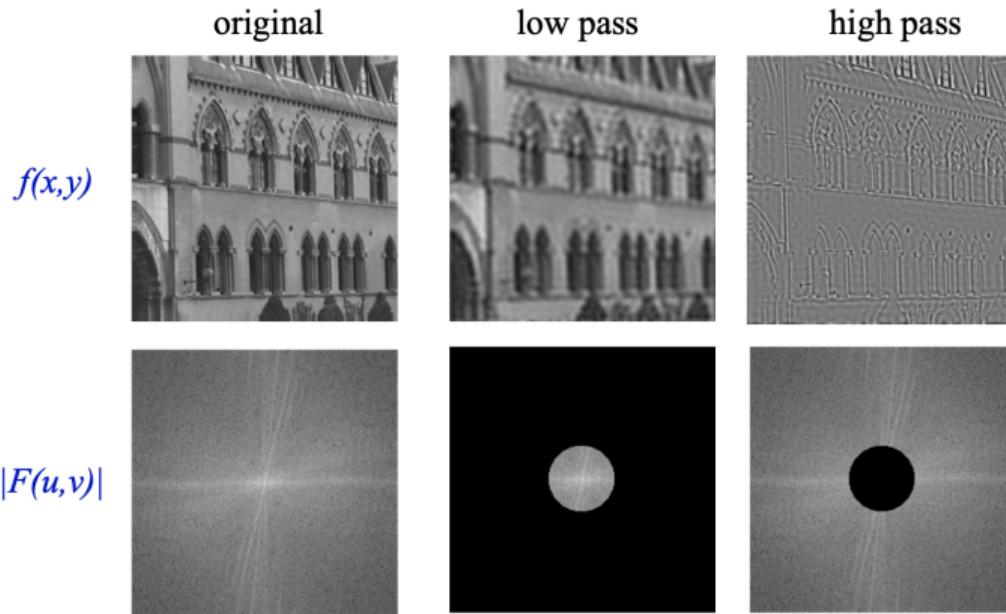
$|F(u,v)|$

Inverse Fourier
transform



$|G(u,v)|$

Filters in space and frequencies



Low-pass and high-pass filtering

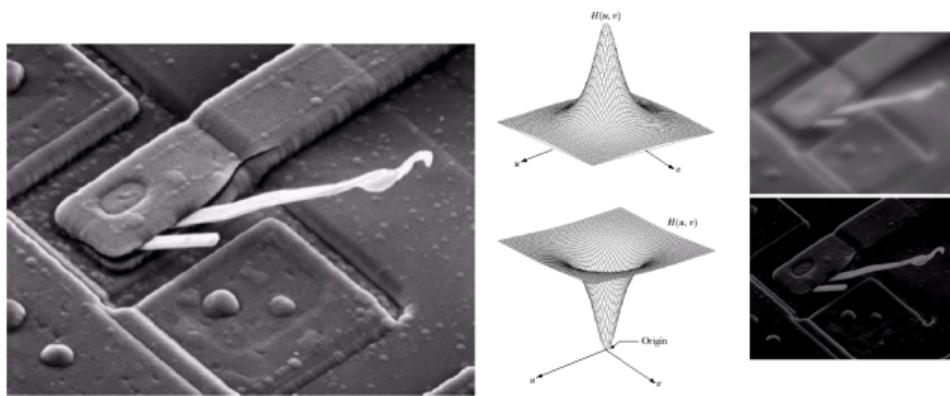


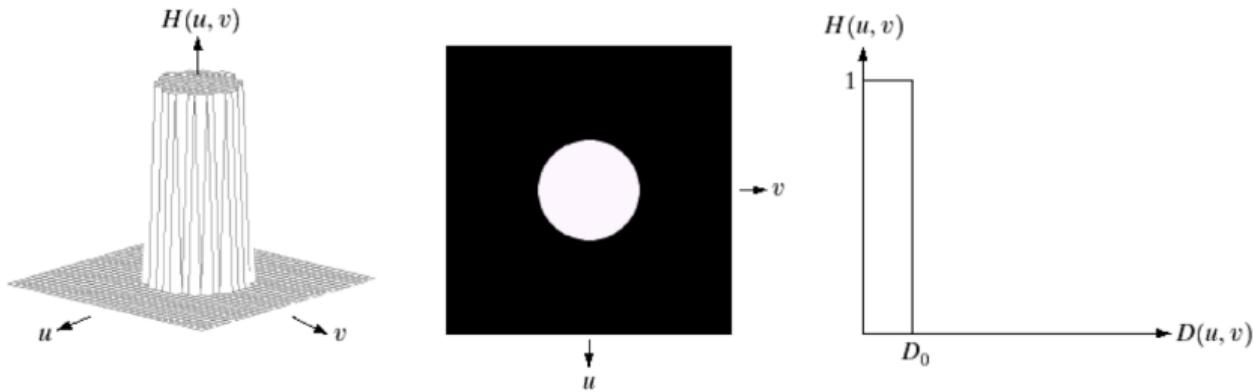
Image smoothing

Smoothing is achieved in the frequency domain by dropping out the high frequency components. This effect is obtained by applying a low-pass filter.
We can take into account different low-pass filters, for instance:

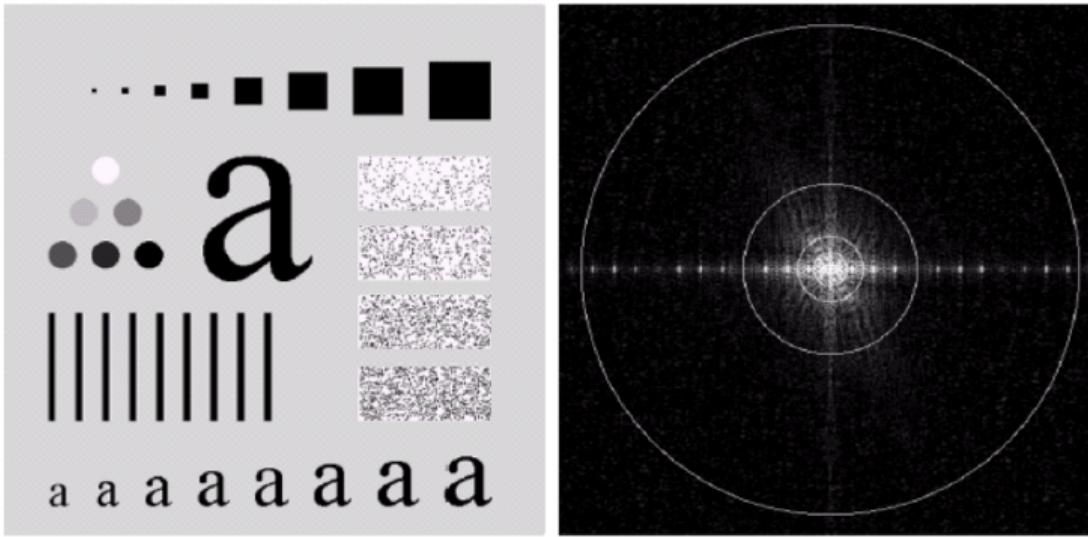
- ▶ Ideal low-pass
- ▶ Butterworth
- ▶ Gaussian

Ideal low-pass filter

Simply cuts off all high frequency components that are a specified distance D_0 from the origin of the transform.

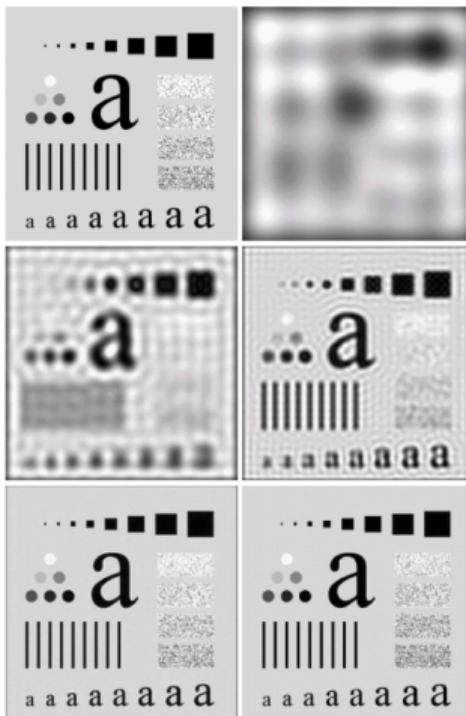


Ideal low-pass filter



Above we show an image, its Fourier spectrum and a series of ideal low pass filters of radius 5, 15, 30, 80 and 230 superimposed on top of it.

Ideal low-pass filter (notice the ringing effects)



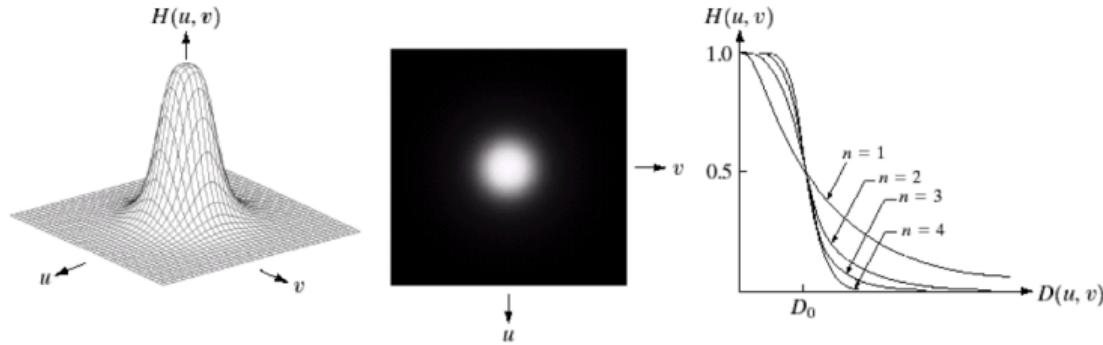
The original image and the five different filtered images with $D_0 = 5, 15, 30, 80, 230$.

From: Digital Image Processing 3rd ed. Gonzalez & Woods

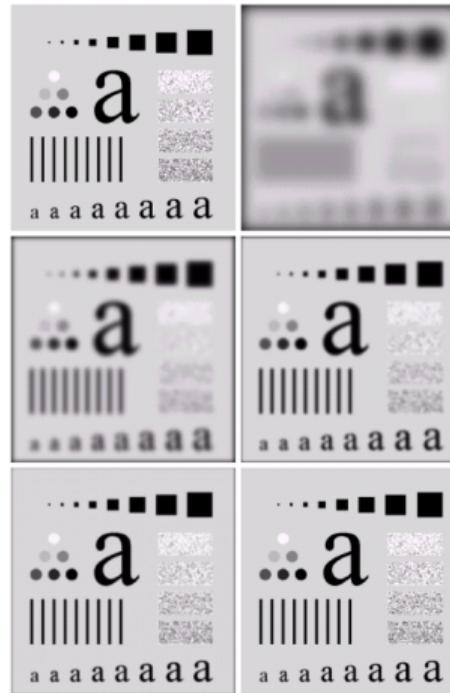
Butterworth low-pass filter

The transfer function of a Butterworth lowpass filter of order n with cutoff frequency at distance D_0 from the origin is defined as:

$$H(u, v) = \frac{1}{1 + [D(u, v)/D_0]^{2n}}$$



Low-pass filtering

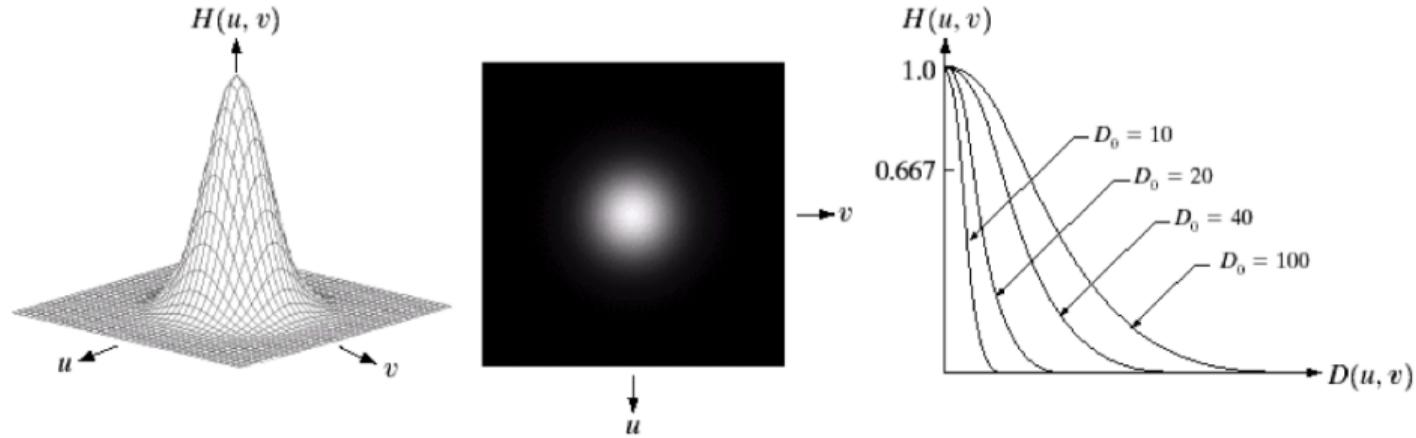


The original image and the five different images filtered by a Butterworth filter of order 2 and $D_0 = 5, 15, 30, 80, 230$.

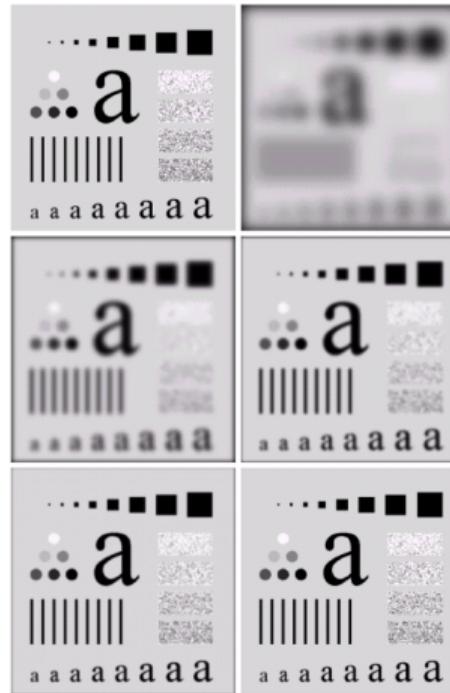
Gaussian low-pass filter

The transfer function of a Gaussian lowpass filter with cutoff frequency at distance D_0 from the origin is defined as:

$$H(u, v) = e^{-\frac{D^2(u, v)}{2D_0^2}}$$

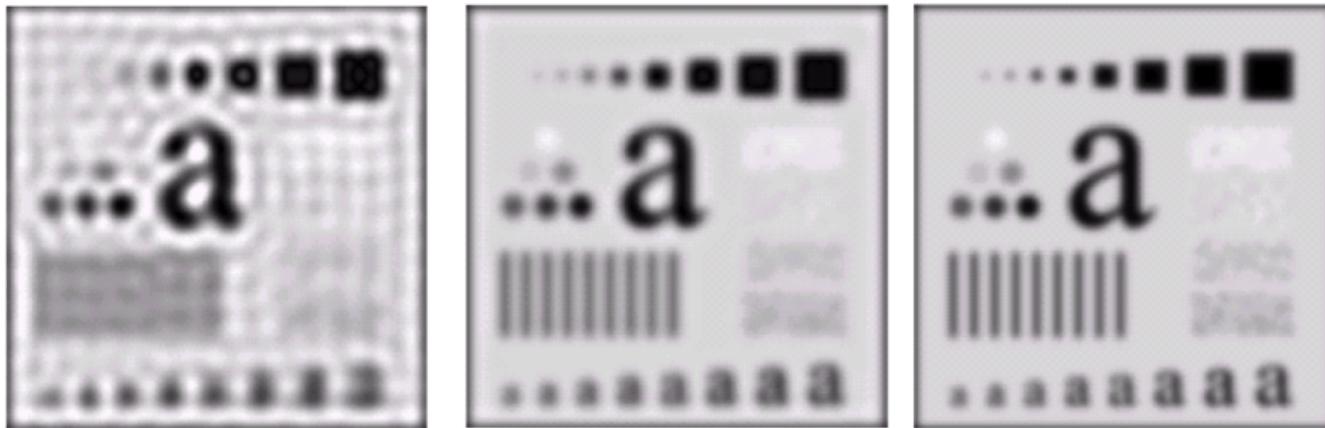


Low-pass filtering



The original image and the five different images filtered by a Gauss filter with
 $D_0 = 5, 15, 30, 80, 230.$

Low-pass filtering comparison

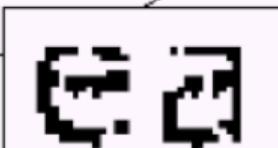


The test image filtered by: (left) Ideal low-pass filter $D_0 = 15$, (center) butterworth low-pass filter $D_0 = 15$, (right) Gaussian low-pass filter $D_0 = 15$

Low-pass filtering application

A low pass Gaussian filter is used to connect broken text or to remove blemishes in a photograph

Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.



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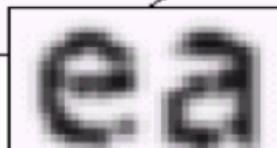
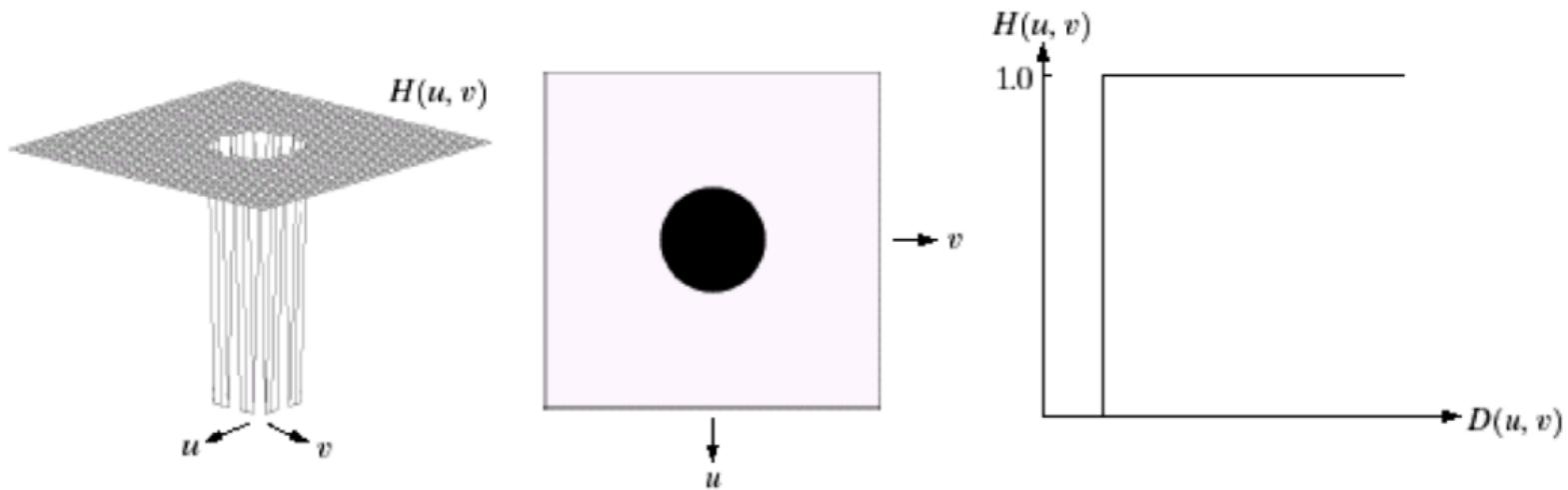


Image enhancement

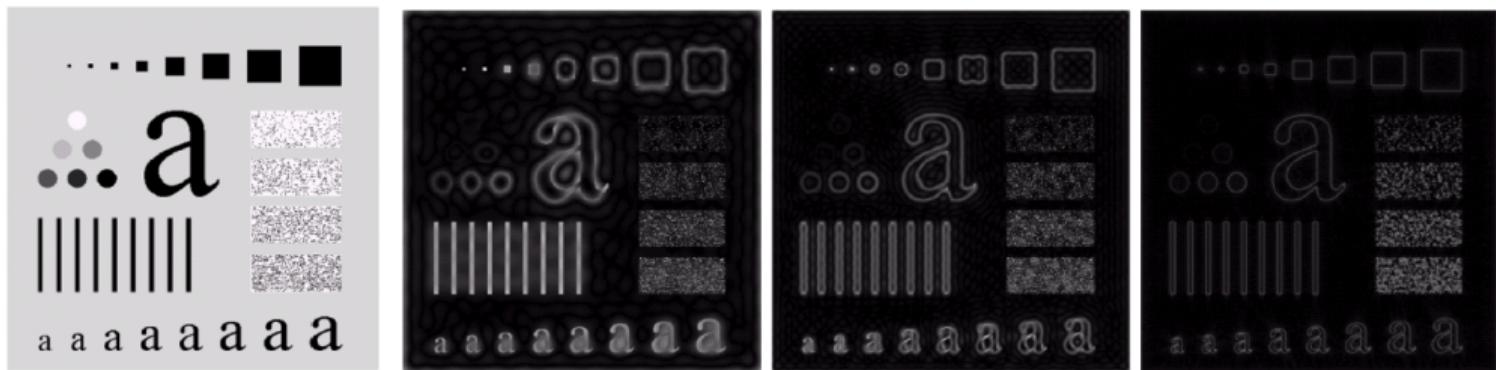
- ▶ Fine details in images are associated with high frequency components
- ▶ High pass filters only pass the high frequencies, dropping the low ones
- ▶ High pass filters are the reverse of low pass filters, so:

$$H_{hp}(u, v) = 1 - H_{lp}(u, v)$$

Ideal High-pass filter



Ideal High-pass filter

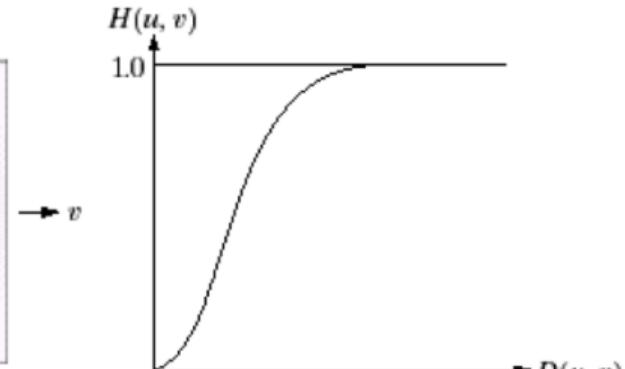
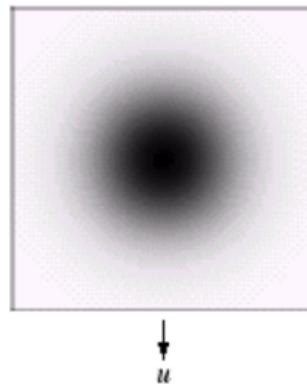
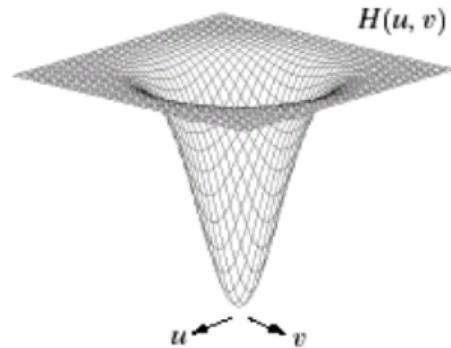


The original image and three different images filtered by an ideal high-pass filter with $D_0 = 15, 30, 80$.

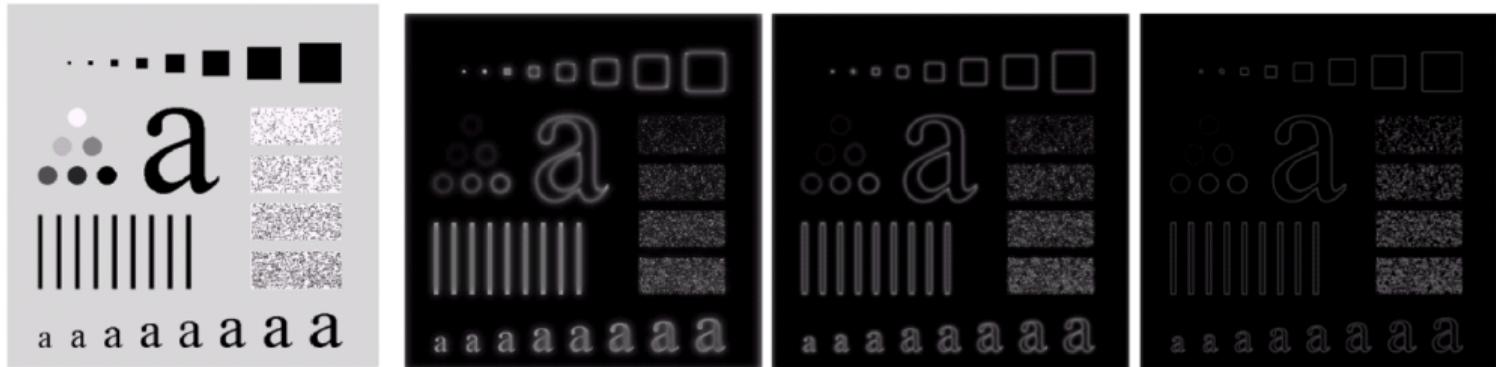
Gaussian high-pass filter

The Gaussian high-pass filter with cutoff frequency at distance D_0 from the origin is defined as:

$$H(u, v) = 1 - e^{-\frac{D^2(u,v)}{2D_0^2}}$$



Gaussian high-pass filtering



The original image and three different images filtered by a Gauss filter with $D_0 = 15, 30, 80$.