

Digital Image Processing

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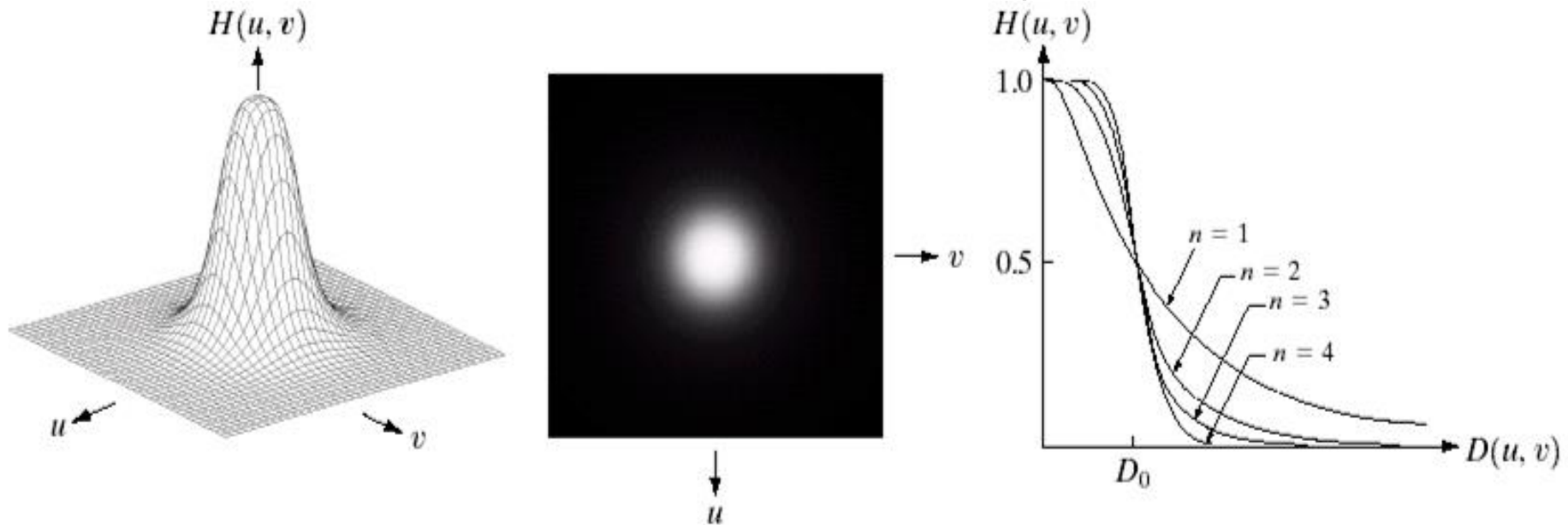
Low Pass Filters in Frequency Domain

- Ideal low pass filters
- Butterworth low pass filters
- Gaussian low pass filters

Butterworth Lowpass Filters

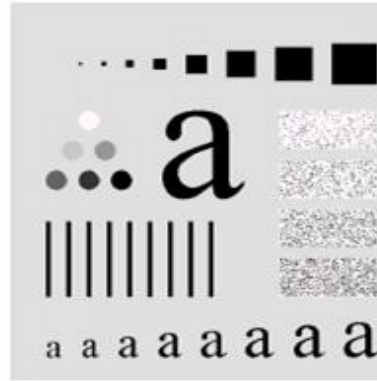
The transfer function of a Butterworth lowpass filter of order n with cutoff frequency at distance D_0 from the origin is defined as:

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



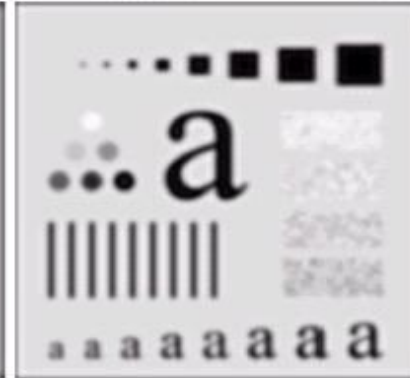
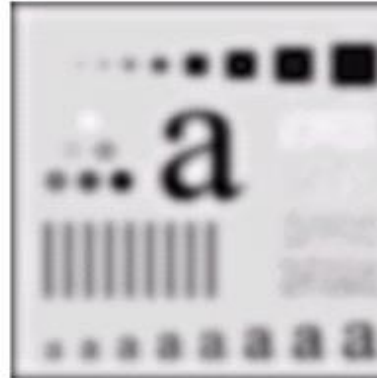
Butterworth Lowpass Filter (cont...)

Original
image



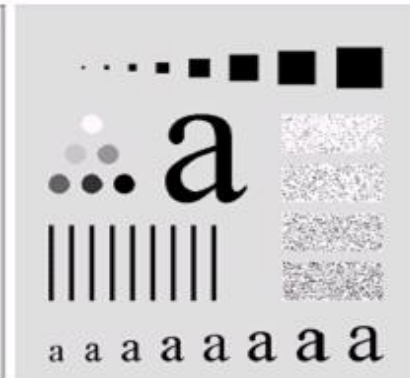
Result of filtering
with Butterworth filter
of order 2 and cutoff
radius 5

Result of filtering with
Butterworth filter of
order 2 and cutoff
radius 15



Result of filtering
with Butterworth
filter of order 2 and
cutoff radius 30

Result of filtering with
Butterworth filter of
order 2 and cutoff
radius 80



Result of filtering
with Butterworth
filter of order 2 and
cutoff radius 230

Butterworth Lowpass Filter (cont...)



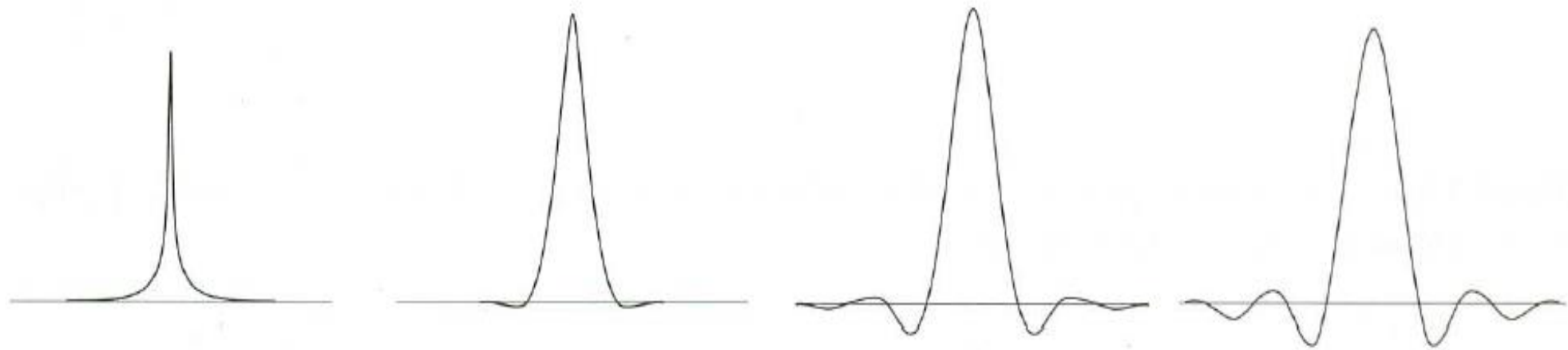
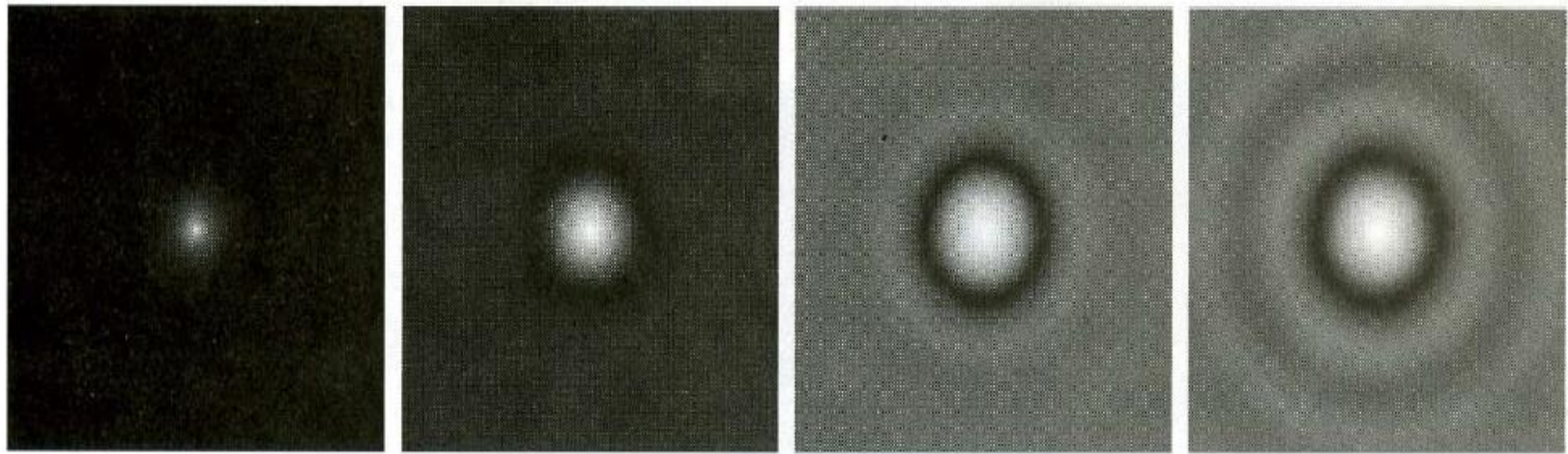
Result of filtering
with Butterworth filter
of order 2 and cutoff
radius 5

Butterworth Lowpass Filter (cont...)

Images taken from Gonzalez & Woods, Digital Image Processing (2002)

Result of filtering with
Butterworth filter of
order 2 and cutoff
radius 15





a b c d

FIGURE 4.16 (a)–(d) Spatial representation of BLPFs of order 1, 2, 5, and 20, and corresponding gray-level profiles through the center of the filters (all filters have a cutoff frequency of 5). Note that ringing increases as a function of filter order.

Calculate the spatial domain to the frequency domain to perform Butterworth Low pass filter, Where cutoff frequency = 20 and $n = 2$

$$F(x, y) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

- Input image $f(x, y)$ in a spatial domain.
- Multiply the input image with $(-1)^{x+y}$ to move the transform in the center.

$$\begin{matrix} x+y \\ (-1) \end{matrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix}$$

Butterworth Low pass filter

- After the transform center arrangement we got the results as follows.

$$\begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{bmatrix}$$

Butterworth Low pass filter: Compute the DFT of the image

- Compute the DFT of the image.
- $F(u, v) = \text{kernel} * f(x, y) * \text{kernel}^{\text{Transpose}}$

$$f(u, v) = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix} \times \begin{pmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix}$$

Butterworth Low pass filter: Compute the DFT of the image

$$\Rightarrow \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 4 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 8 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Distance calculation

- Distance formula = $(x^2 + y^2)^{1/2}$

$$V = \begin{bmatrix} -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \end{bmatrix} \quad U = \begin{bmatrix} -2 & -2 & -2 & -2 \\ -1 & -1 & -1 & -1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$$D(U, V) = \begin{bmatrix} (-2, -2) & (-1, -2) & (0, -2) & (1, -2) \\ (-2, -1) & (-1, -1) & (0, -1) & (1, -1) \\ (-2, 0) & (-1, 0) & (0, 0) & (1, 0) \\ (-2, 1) & (-1, 1) & (0, 1) & (1, 1) \end{bmatrix}$$

Distance calculation

- $D(-2, -2) = 2.82$
- $D(-1, -2) = 2.23$
- $D(0, -2) = 2$
- $D(1, -1) = 2.23$
- $D(-2, -1) = 2.23$
- $D(-1, -1) = 1.41$
- $D(0, -1) = 1$
- $D(1, -1) = 1.41$
- $D(-2, 0) = 2$
- $D(-1, 0) = 1.41$
- $D(0, 0) = 0$
- $D(1, 0) = 1$
- $D(-2, 1) = 2.23$
- $D(-1, 1) = 1.41$
- $D(0, 1) = 1$
- $D(1, 1) = 1.41$

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

$$D(U, V) = \begin{bmatrix} 2.8284 & 2.2361 & 2.0000 & 2.2361 \\ 2.2361 & 1.4142 & 1.0000 & 1.4142 \\ 2.0000 & 1.0000 & 0 & 1.0000 \\ 2.2361 & 1.4142 & 1.0000 & 1.4142 \end{bmatrix}$$

Apply Butterworth formula

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

$$\text{Butterworth} = \begin{pmatrix} 0.99 & 0.99 & 0.99 & 0.99 \\ 0.99 & 1 & 1 & 1 \\ 0.99 & 1 & 1 & 1 \\ 0.99 & 1 & 1 & 1 \end{pmatrix}$$

Applying the Butterworth filter to the image
in Frequency domain

$$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 8 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 0.99 & 0.99 & 0.99 & 0.99 \\ 0.99 & 1 & 1 & 1 \\ 0.99 & 1 & 1 & 1 \\ 0.99 & 1 & 1 & 1 \end{bmatrix}$$

Inverse Fourier transform

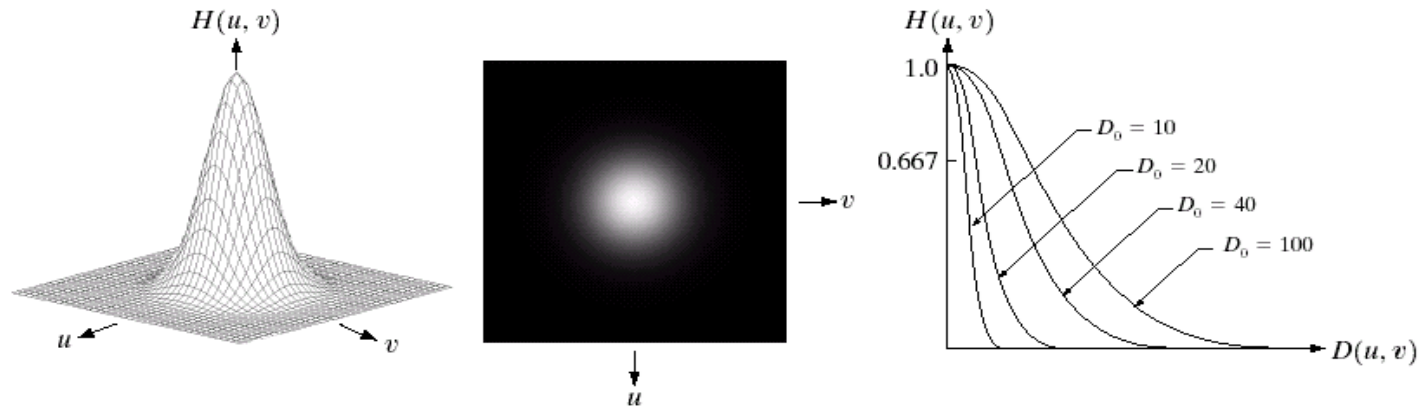
$$f(x,y) = \frac{1}{16} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{pmatrix} \times \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 7.9101 & 0 & 8.0000 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{pmatrix}$$

$$f(x,y) = \begin{pmatrix} 0.9944 & -0.0056 & 0.9944 & -0.0056 \\ -0.9944 & 0.0056 & -0.9944 & 0.0056 \\ 0.9944 & -0.0056 & 0.9944 & -0.0056 \\ -0.9944 & 0.0056 & -0.9944 & 0.0056 \end{pmatrix}$$

Gaussian Lowpass Filters

The transfer function of a Gaussian lowpass filter is defined as:

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



Calculate the spatial domain to the frequency domain to perform the Gaussian Low pass filter, Where cutoff frequency = 1

$$F(x, y) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

- Input image $f(x, y)$ in a spatial domain.
- Multiply the input image with $(-1)^{x+y}$ to move the transform in the center.

$$(-1)^{x+y} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix}$$

Gaussian Low pass filter

- After the transform center arrangement we got the results as follows.

$$\begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{bmatrix}$$

Gaussian Low pass filter: Compute the DFT of the image

- Compute the DFT of the image.
- $F(u, v) = \text{kernel} * f(x, y) * \text{kernel}^T$

$$f(u, v) = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix} \times \begin{pmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix}$$

Gaussian Low pass filter: Compute the DFT of the image

$$\Rightarrow \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 4 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 8 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Gaussian Low pass filter formula

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

$$X = \begin{bmatrix} -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \end{bmatrix} \quad Y = \begin{bmatrix} -2 & -2 & -2 & -2 \\ -1 & -1 & -1 & -1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

```
distances = sqrt(x.^2 + y.^2);
```

$$\begin{bmatrix} 2.8284 & 2.2361 & 2.0000 & 2.2361 \\ 2.2361 & 1.4142 & 1.0000 & 1.4142 \\ 2.0000 & 1.0000 & 0 & 1.0000 \\ 2.2361 & 1.4142 & 1.0000 & 1.4142 \end{bmatrix}$$

```
gaussian_filter = exp(-(distances)2 / (2*sigma^2));
```

```
gaussian_filter =
```

```

$$\begin{bmatrix} 0.0183 & 0.0821 & 0.1353 & 0.0821 \\ 0.0821 & 0.3679 & 0.6065 & 0.3679 \\ 0.1353 & 0.6065 & 1.0000 & 0.6065 \\ 0.0821 & 0.3679 & 0.6065 & 0.3679 \end{bmatrix}$$

```

```
filtered_image_F = F (u v) .* gaussian_filter;
```

```
filtered_image_F =
```

```

$$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1.0827 & 0 & 8.0000 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

```

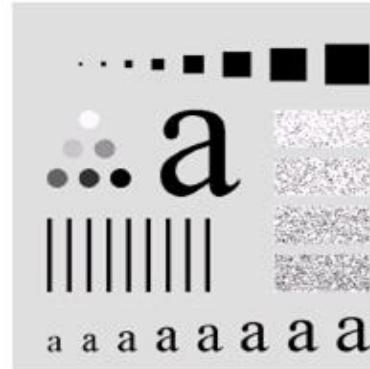
Applying inverse Fourier transform

$$f(x, y) = \frac{1}{16} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{pmatrix} \times \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1.0827 & 0 & 8.0000 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{pmatrix}$$

$$f(x, y) = \begin{pmatrix} 0.5677 & -0.4323 & 0.5677 & -0.4323 \\ -0.5677 & 0.4323 & -0.5677 & 0.4323 \\ 0.5677 & -0.4323 & 0.5677 & -0.4323 \\ -0.5677 & 0.4323 & -0.5677 & 0.4323 \end{pmatrix}$$

Gaussian Lowpass Filters (cont...)

Original
image



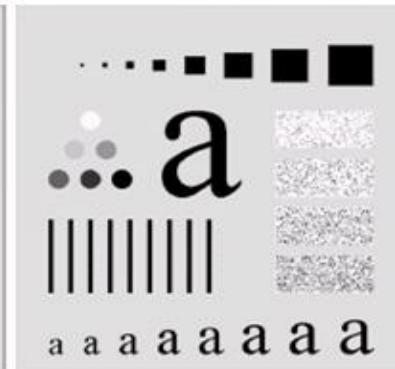
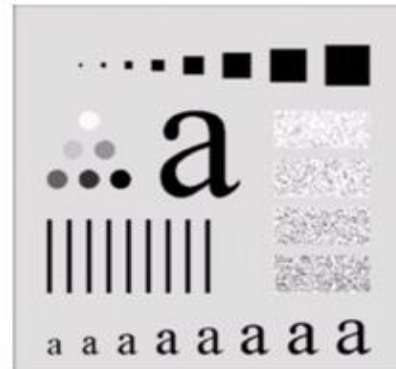
Result of filtering
with Gaussian filter
with cutoff radius 5

Result of filtering
with Gaussian
filter with cutoff
radius 15



Result of filtering
with Gaussian filter
with cutoff radius 30

Result of filtering
with Gaussian
filter with cutoff
radius 85

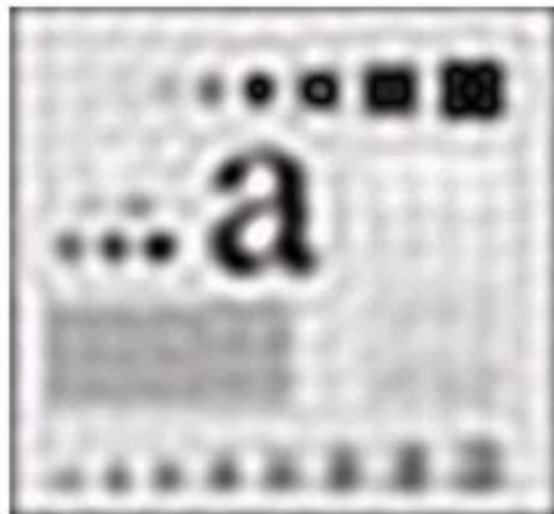


Result of filtering
with Gaussian filter
with cutoff radius
230

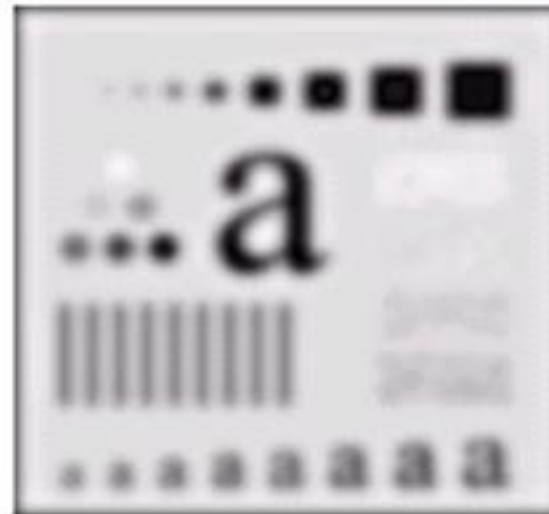
Lowpass Filters Compared

Images taken from Gonzalez & Woods, Digital Image Processing (2002)

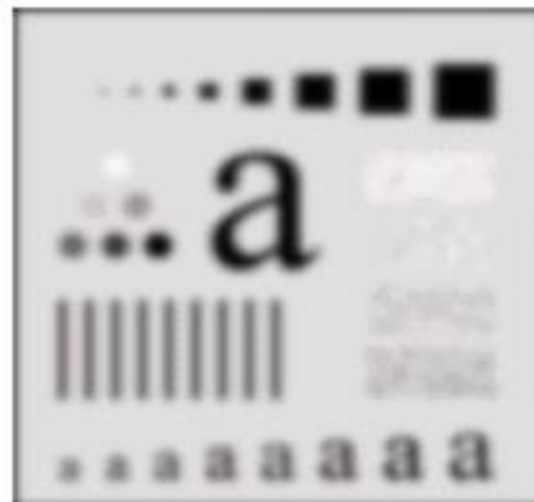
Result of filtering
with ideal low pass
filter of radius 15



Result of filtering
with Butterworth
filter of order 2
and cutoff radius
15

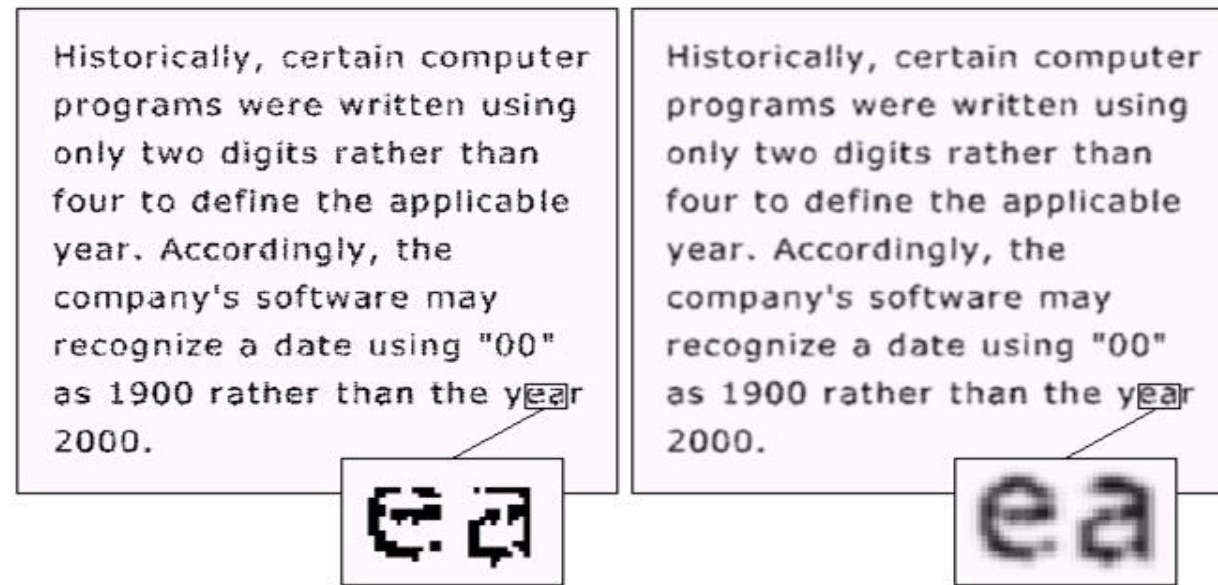


Result of filtering
with Gaussian
filter with cutoff
radius 15



Lowpass Filtering Examples

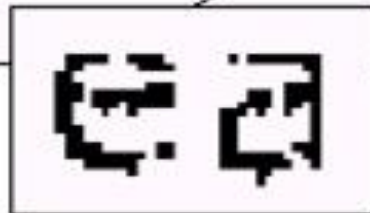
A low pass Gaussian filter is used to connect broken text



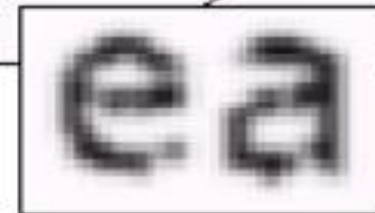
Lowpass Filtering Examples

Images taken from Gonzalez & Woods, Digital Image Processing (2002)

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.



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.



Lowpass Filtering Examples (cont...)

Different low pass Gaussian filters are used to remove blemishes in a photograph



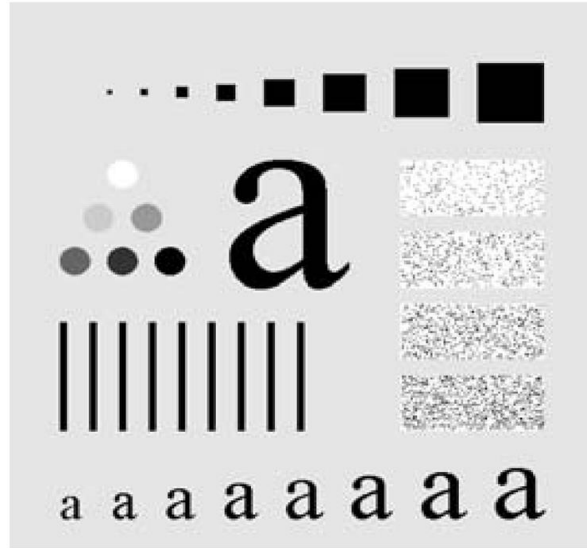
Lowpass Filtering Examples (cont...)

Images taken from Gonzalez & Woods, Digital Image Processing (2002)

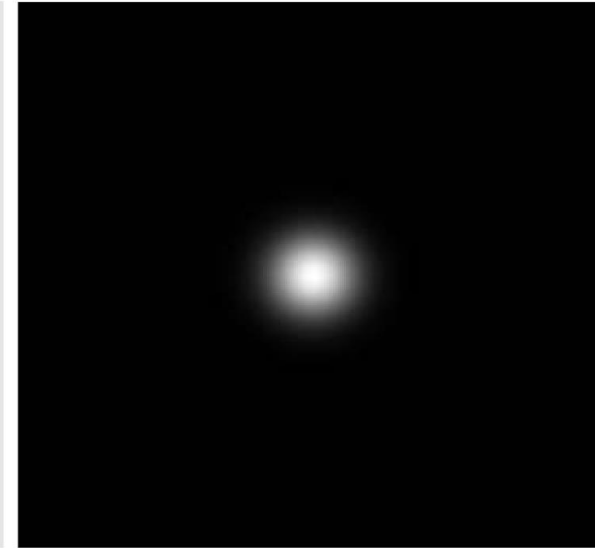


Lowpass Filtering Examples (cont...)

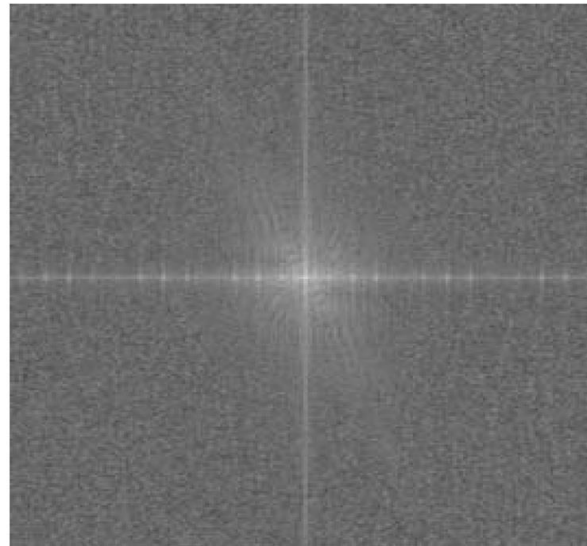
Original image



Gaussian lowpass filter



Spectrum of original image



Processed image



Sharpening in the Frequency Domain

Edges and fine detail in images are associated with high frequency components

High pass filters – only pass the high frequencies, drop the low ones

High pass filters are precisely the reverse of low pass filters, so:

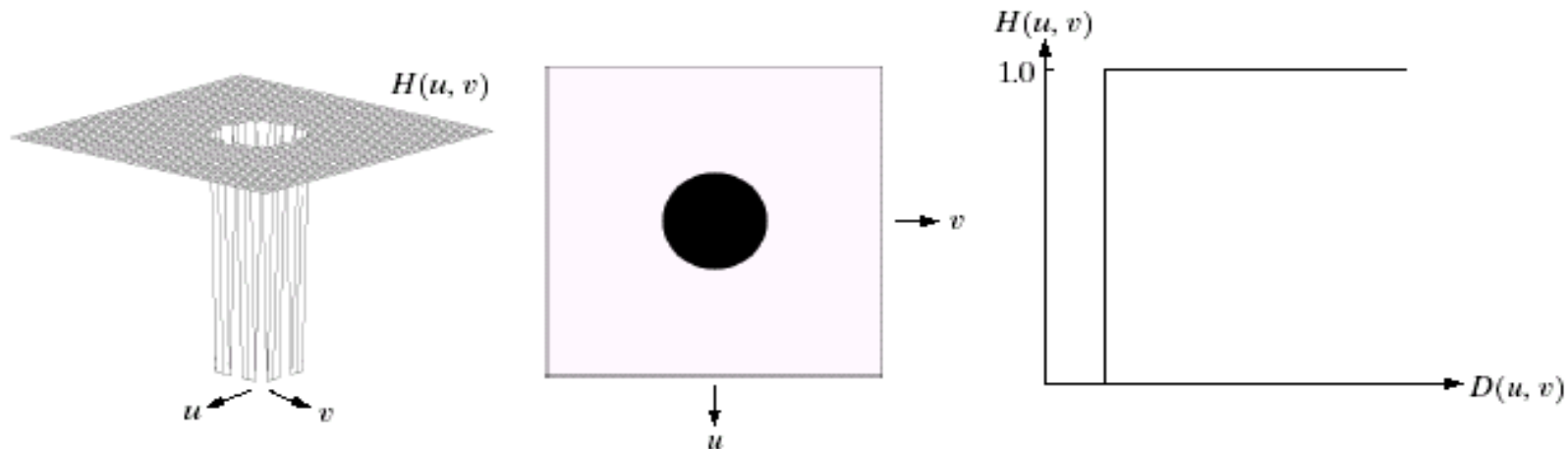
$$H_{hp}(u, v) = 1 - H_{lp}(u, v)$$

Ideal High Pass Filters

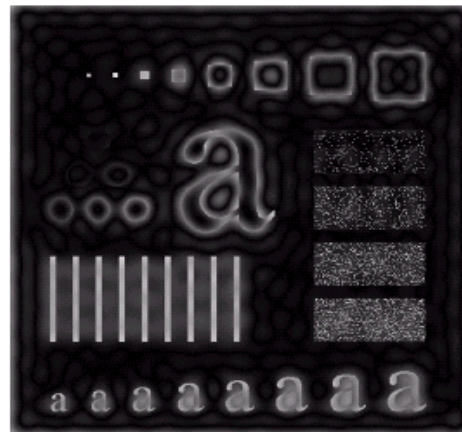
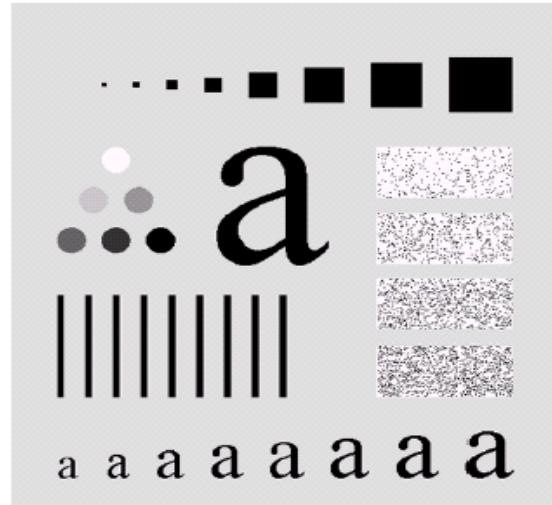
The ideal high pass filter is given as:

$$H(u, v) = \begin{cases} 0 & \text{if } D(u, v) \leq D_0 \\ 1 & \text{if } D(u, v) > D_0 \end{cases}$$

where D_0 is the cut off distance as before



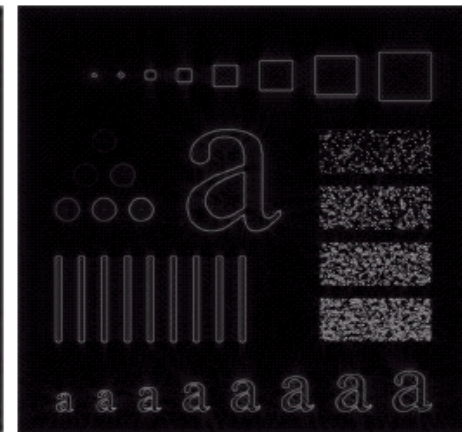
Ideal High Pass Filters (cont...)



Results of ideal
high pass filtering
with $D_0 = 15$



Results of ideal
high pass filtering
with $D_0 = 30$



Results of ideal
high pass filtering
with $D_0 = 80$

Calculate the spatial domain to the frequency domain to perform Ideal High pass filter, Where cutoff frequency = 0.5

$$F(x, y) = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

- Input image $f(x, y)$ in a spatial domain.
- Multiply the input image with $(-1)^{x+y}$ to move the transform in the center.

$$(-1)^{x+y} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix}$$

Ideal Low pass filter

- After the transform center arrangement we got the results as follows.

$$\begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{bmatrix}$$

Ideal Low pass filter: Compute the DFT of the image

- Compute the DFT of the image.
- $F(u, v) = \text{kernel} * f(x, y) * \text{kernel}^T$

$$f(u, v) = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix} \times \begin{pmatrix} 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \\ 1 & 0 & 1 & 0 \\ -1 & 0 & -1 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix}$$

Ideal Low pass filter: Compute the DFT of the image

$$\Rightarrow \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 4 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 8 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Distance measure

$$U = \begin{bmatrix} -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \end{bmatrix}$$

$$V = \begin{bmatrix} -2 & -2 & -2 & -2 \\ -1 & -1 & -1 & -1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$$D(U, V) = \begin{bmatrix} (-2, -2) & (-1, -2) & (0, -2) & (1, -2) \\ (-2, -1) & (-1, -1) & (0, -1) & (1, -1) \\ (-2, 0) & (-1, 0) & (0, 0) & (1, 0) \\ (-2, 1) & (-1, 1) & (0, 1) & (1, 1) \end{bmatrix}$$

Ideal Low pass filter:

- Now compute the distances
- $D(u, v) = \sqrt{u.^2 + v.^2}$
- $D(-2, -2) = (-2)^2 + (-2)^2 = (8)^{1/2} = 2.82$
- $D(-2, -1) = (-2)^2 + (-1)^2 = (5)^{1/2} = 2.23$
- $D(0, 2) = (0)^2 + (-2)^2 = (4)^{1/2} = 2$
- $D(1, -2) = (1)^2 + (-2)^2 = 2.23$
- $D(-2, -1) = (-2)^2 + (-1)^2 = 1.14$
- $D(-1, -1) = (-1)^2 + (-1)^2 = 2.23$

Ideal Low pass filter: Distance measure

- $D(0, -1) = (0)^2 + (-1)^2 = 1$
- $D(1, -1) = (1)^2 + (-1)^2 = 1.41$
- $D(-2, 0) = (-2)^2 + (0)^2 = 2$
- $D(-1, 0) = (-1)^2 + (0)^2 = 1$
- $D(0, 0) = (0)^2 + (0)^2 = 0$
- $D(1, 0) = (1)^2 + (0)^2 = 1$
- $D(-2, 1) = (-2)^2 + (1)^2 = 2.23$
- $D(-1, 1) = (-1)^2 + (1)^2 = 1.41$
- $D(0, 1) = (0)^2 + (1)^2 = 1$
- $D(1, 1) = (1)^2 + (1)^2 = 1.41$

Ideal Low pass filter: Distance measure

$$D(u,v) = \begin{bmatrix} 2.8284 & 2.2361 & 2 & 2.2361 \\ 2.2361 & 1.4142 & 1 & 1.4142 \\ 2 & 1 & 0 & 1 \\ 2.2361 & 1.4142 & 1 & 1.4142 \end{bmatrix} = \begin{bmatrix} (-2, -2) & (-1, -2) & (0, -2) & (1, -2) \\ (-2, -1) & (-1, -1) & (0, -1) & (1, -1) \\ (-2, 0) & (-1, 0) & (0, 0) & (1, 0) \\ (-2, 1) & (-1, 1) & (0, 1) & (1, 1) \end{bmatrix}$$

Cut off frequency $D_0 = 0.5$

- Distance matrix will be

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \times \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 8 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Cut off frequency $D0 = 0.5$

$$f(x, y) = \frac{1}{16} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} \times \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 8 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix}$$

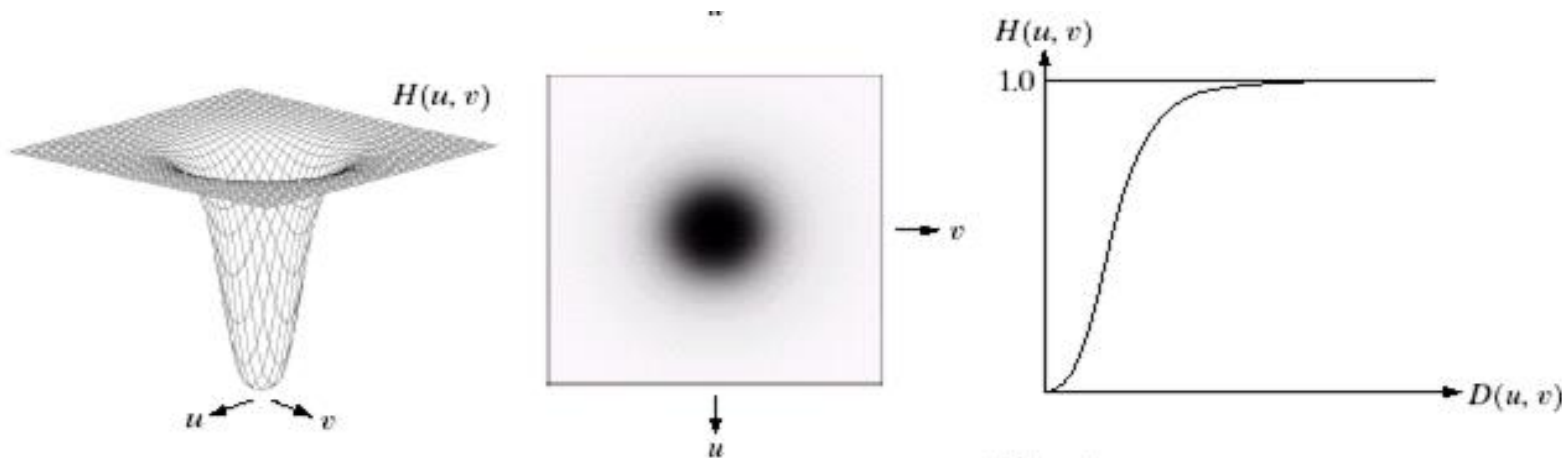
$$f(x, y) = \frac{1}{16} \begin{bmatrix} 8 & 8 & 8 & 8 \\ -8 & -8 & -8 & -8 \\ 8 & 8 & 8 & 8 \\ -8 & -8 & -8 & -8 \end{bmatrix}$$

Butterworth High Pass Filters

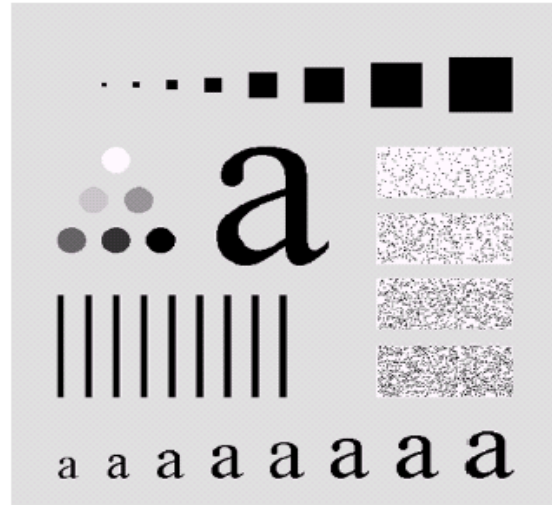
The Butterworth high pass filter is given as:

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

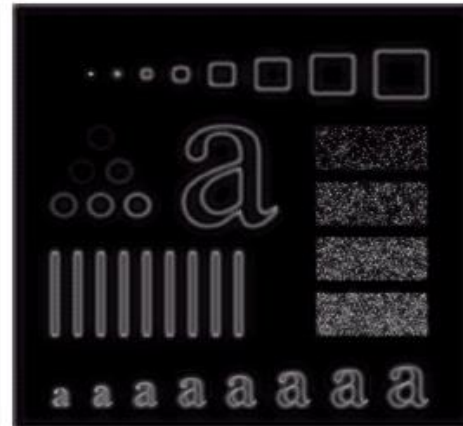
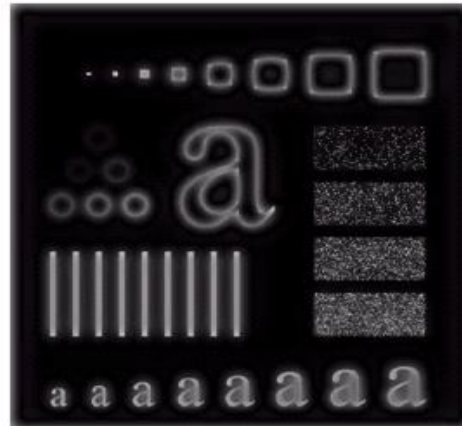
where n is the order and D_0 is the cut off distance as before



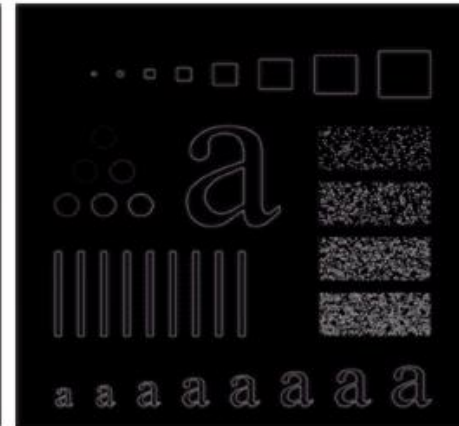
Butterworth High Pass Filters (cont...)



Results of
Butterworth
high pass
filtering of
order 2 with
 $D_0 = 15$



Results of Butterworth high pass
filtering of order 2 with $D_0 = 30$



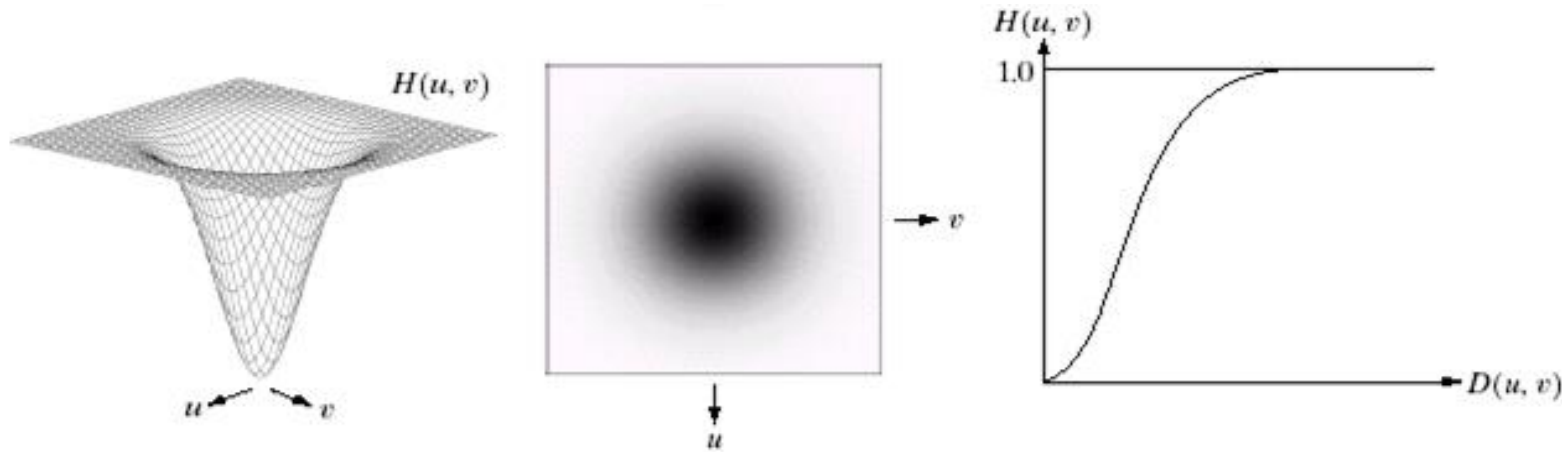
Results of
Butterworth
high pass
filtering of
order 2 with
 $D_0 = 80$

Gaussian High Pass Filters

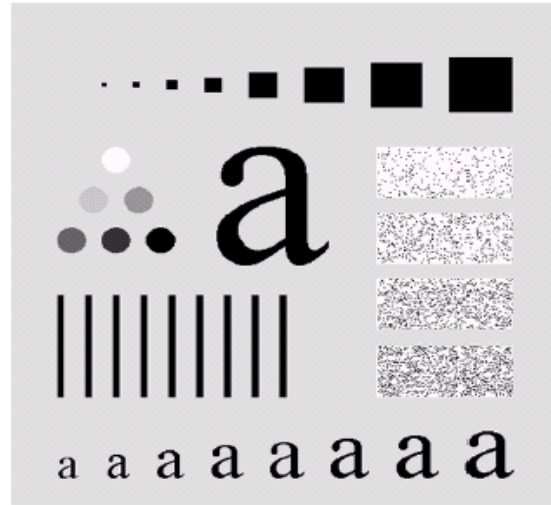
The Gaussian high pass filter is given as:

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

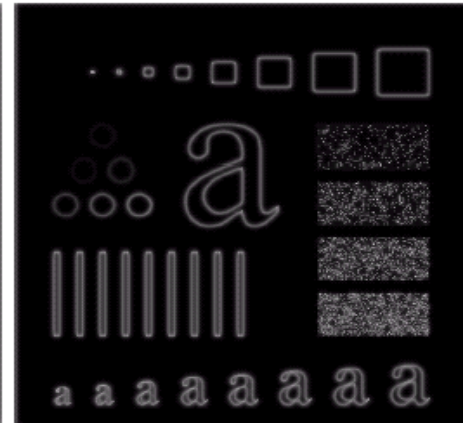
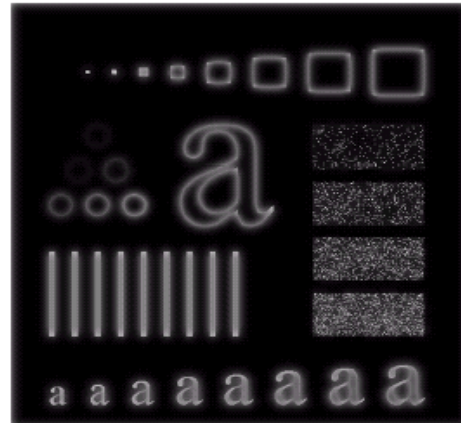
where D_0 is the cut off distance as before



Gaussian High Pass Filters (cont...)



Results of
Gaussian
high pass
filtering with
 $D_0 = 15$

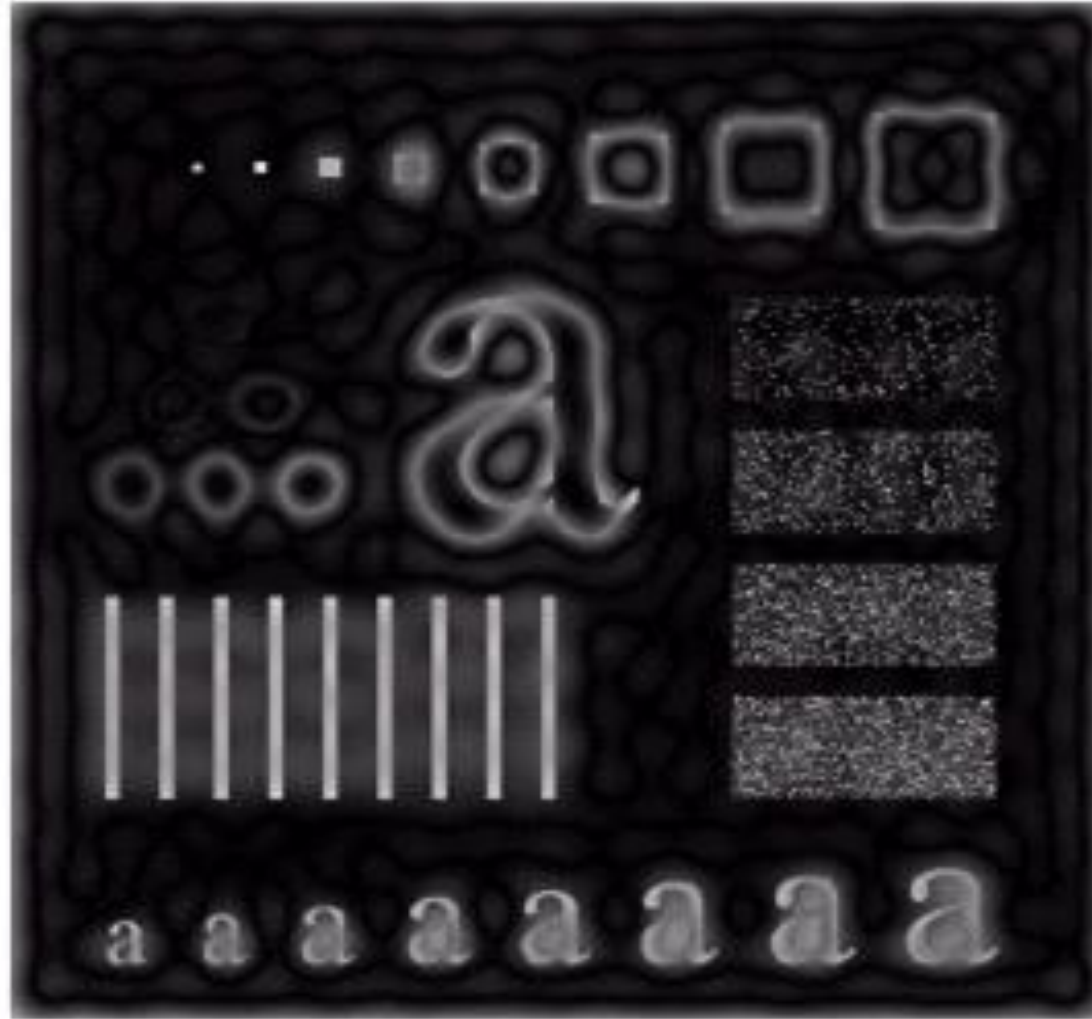


Results of Gaussian high pass
filtering with $D_0 = 30$



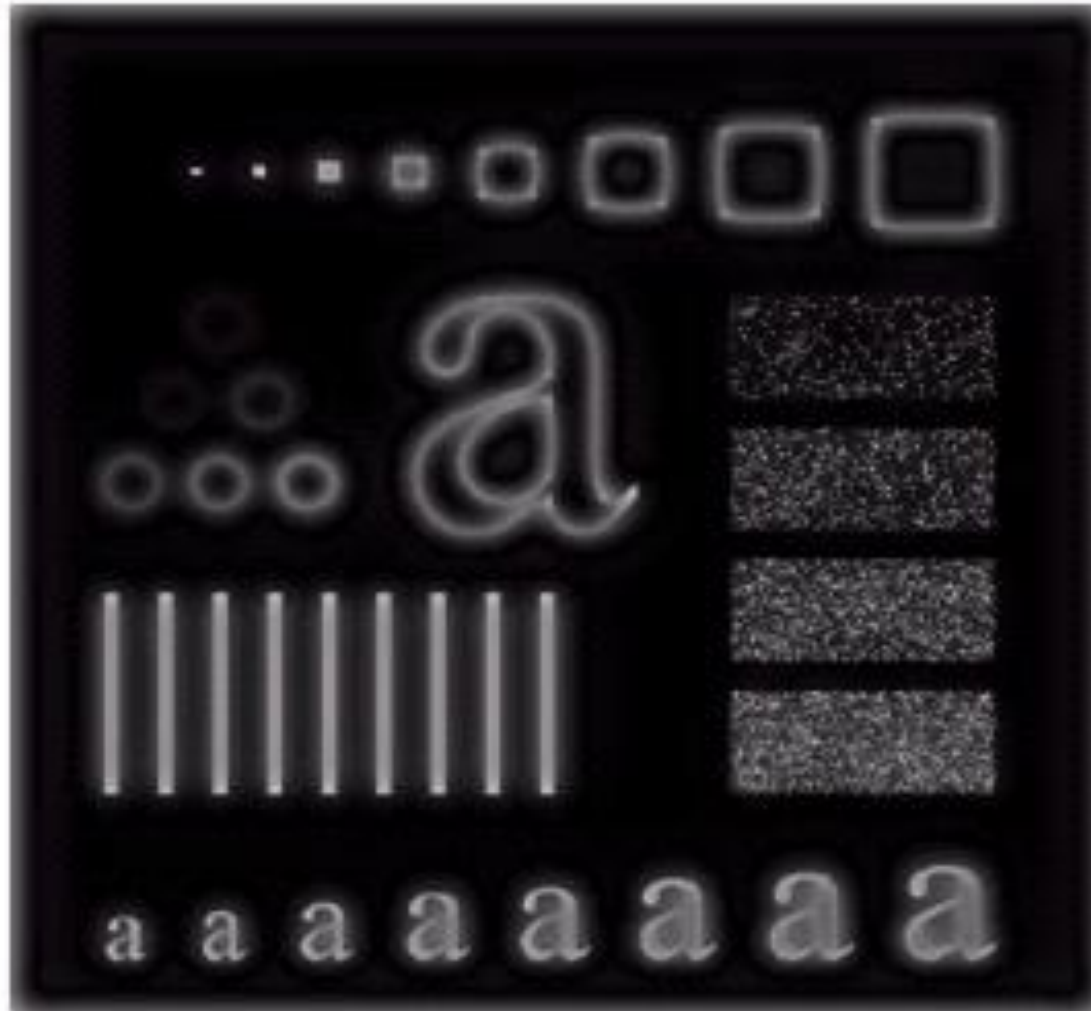
Results of
Gaussian
high pass
filtering with
 $D_0 = 80$

Highpass Filter Comparison



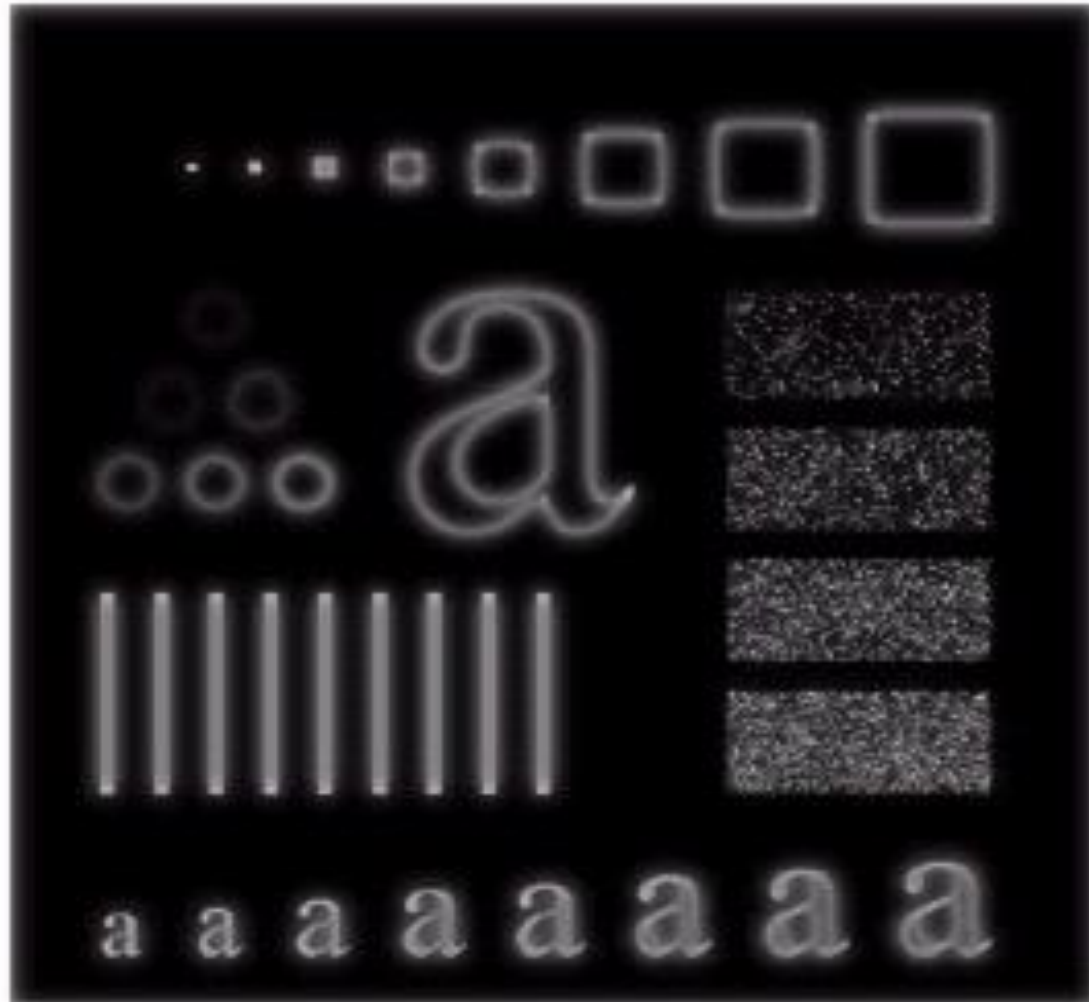
Results of ideal
high pass filtering
with $D_0 = 15$

Highpass Filter Comparison



Results of Butterworth
high pass filtering of order
2 with $D_0 = 15$

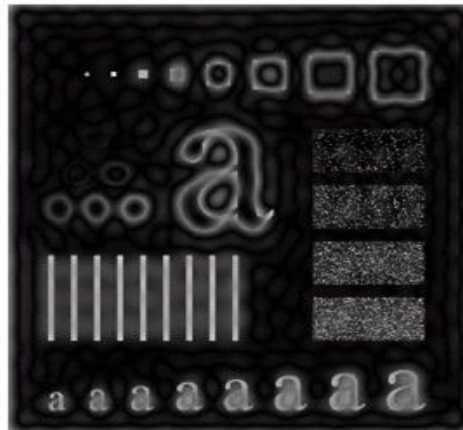
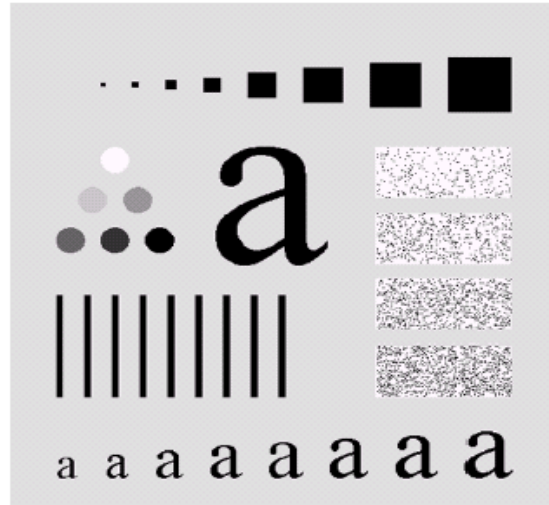
Highpass Filter Comparison



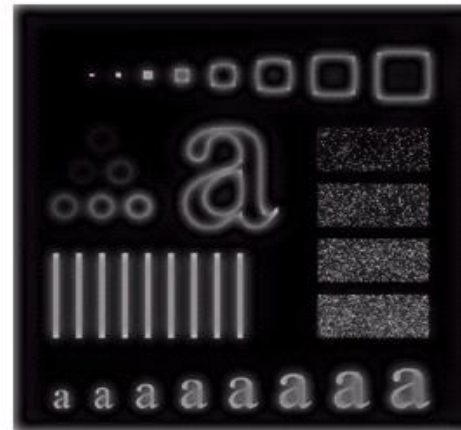
Results of Gaussian
high pass filtering with
 $D_0 = 15$

Highpass Filter Comparison

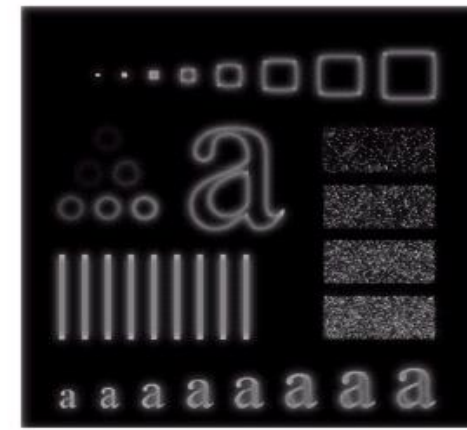
Images taken from Gonzalez & Woods, Digital Image Processing (2002)



Results of ideal
high pass filtering
with $D_0 = 15$

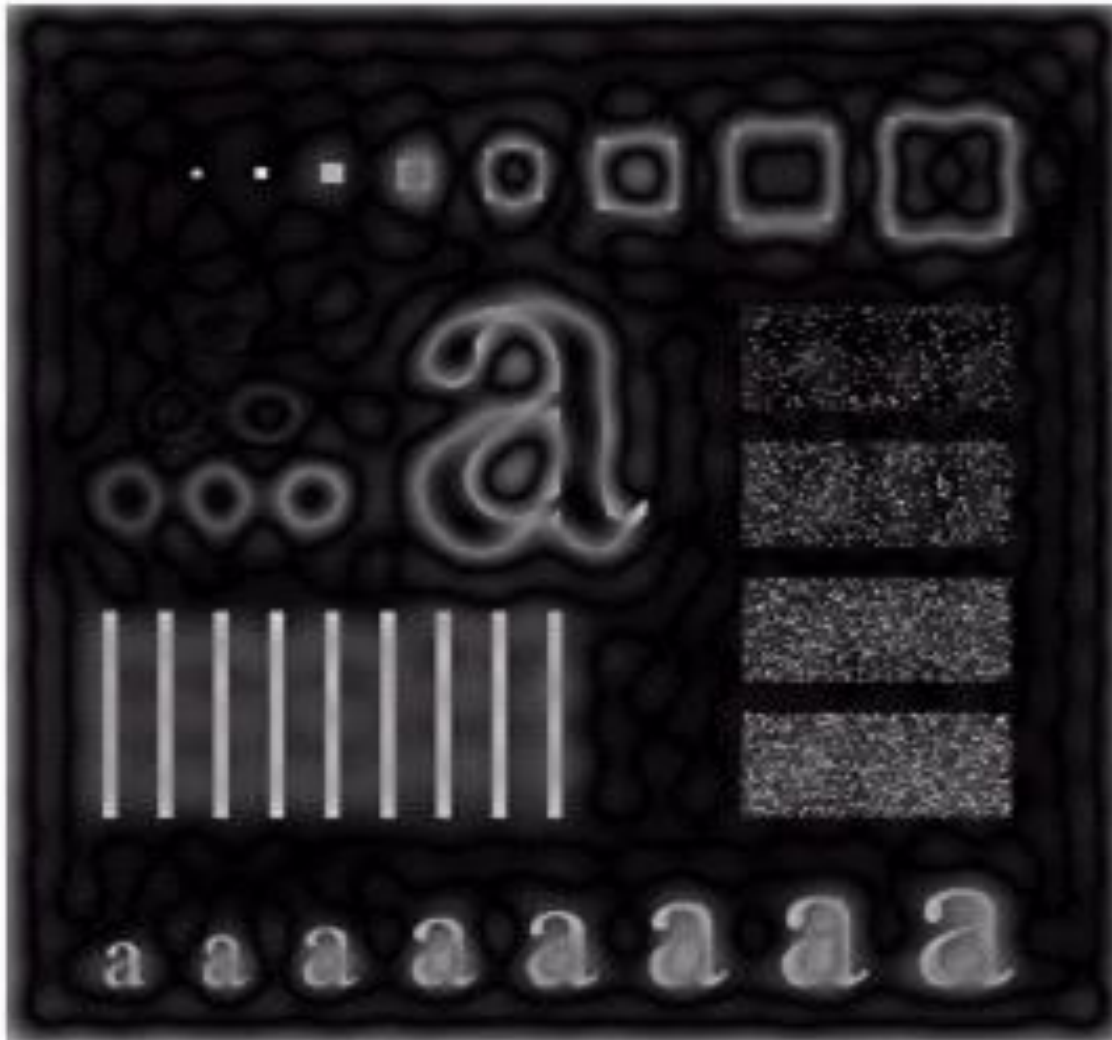


Results of Butterworth
high pass filtering of order
2 with $D_0 = 15$



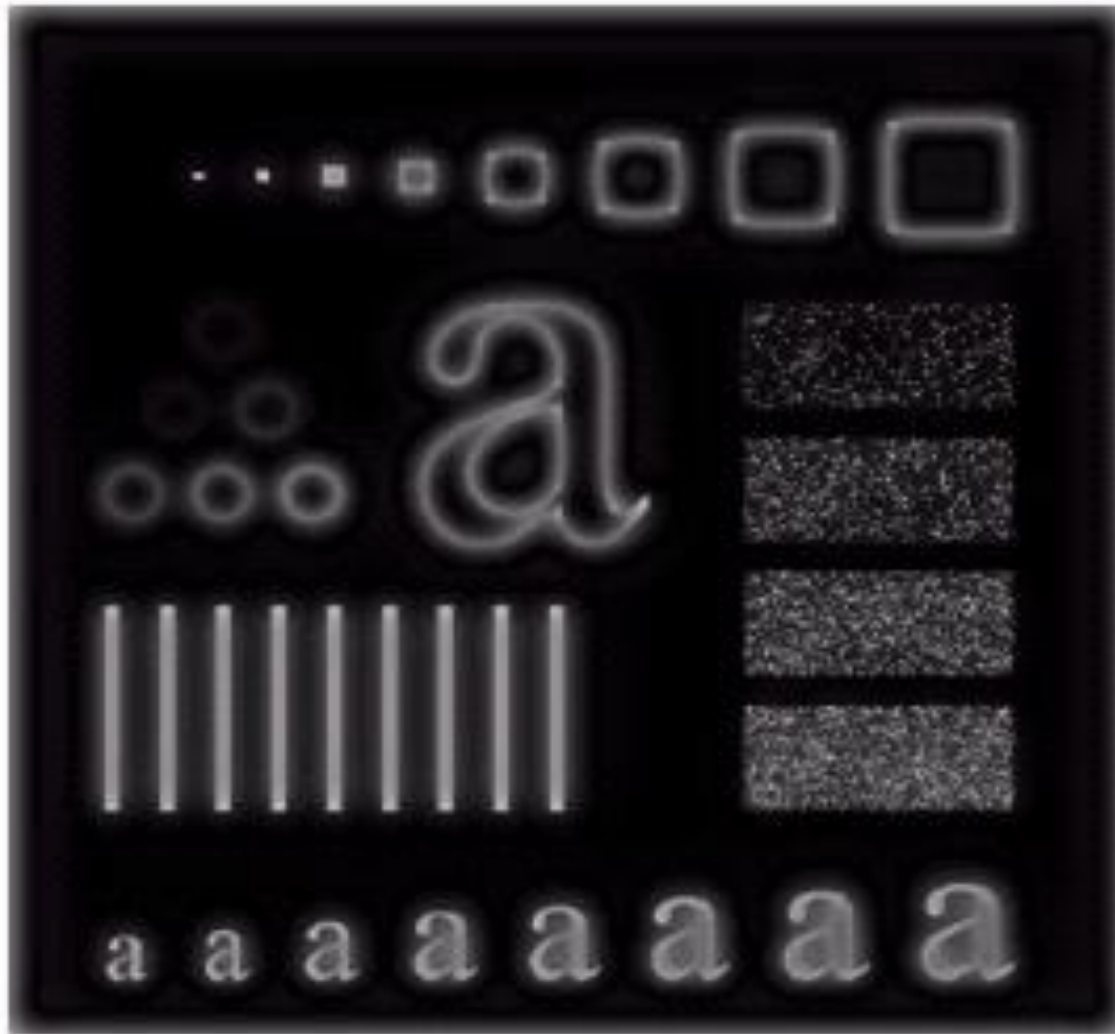
Results of Gaussian
high pass filtering with
 $D_0 = 15$

Highpass Filter Comparison



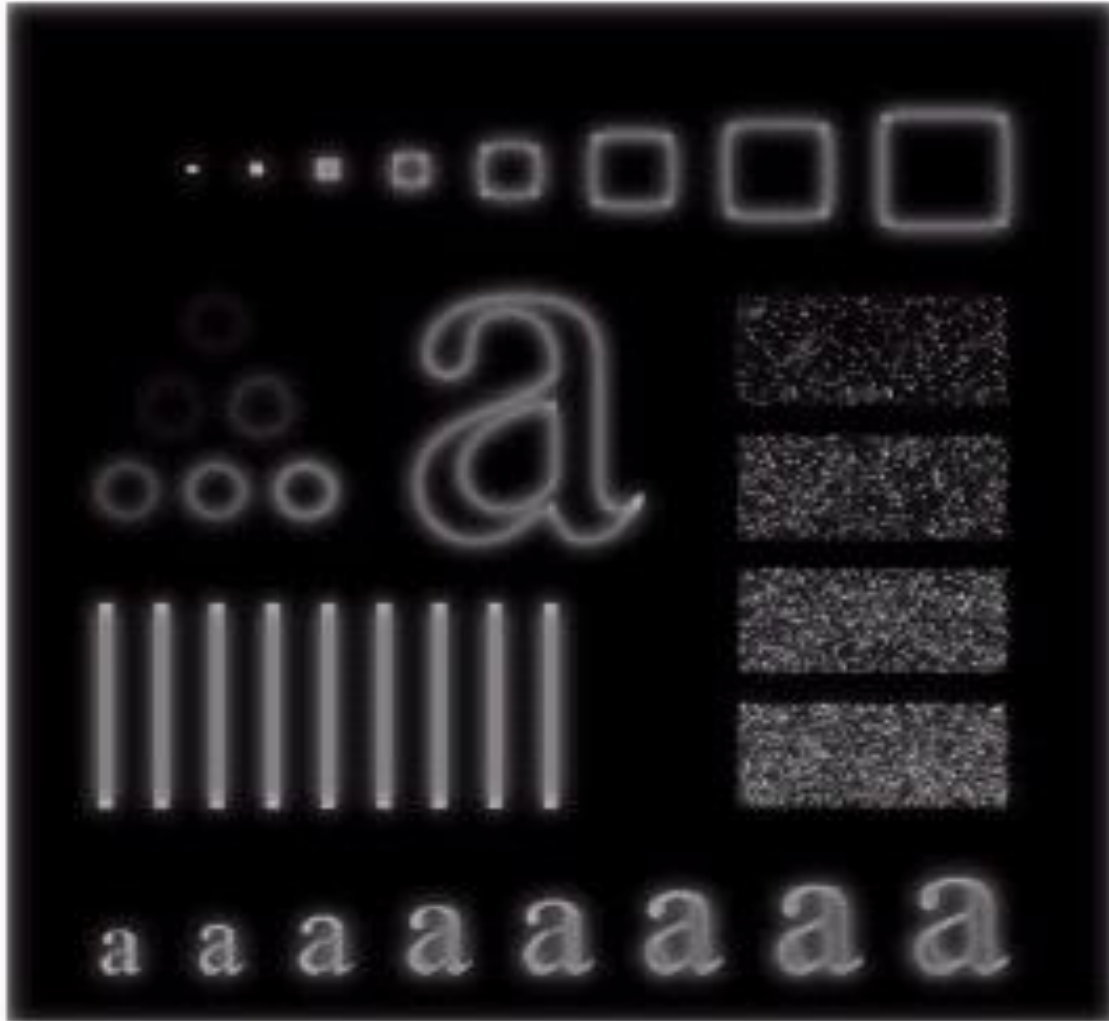
Results of ideal
high pass filtering
with $D_0 = 15$

Highpass Filter Comparison



Results of Butterworth
high pass filtering of order
2 with $D_0 = 15$

Highpass Filter Comparison



Results of Gaussian
high pass filtering with
 $D_0 = 15$

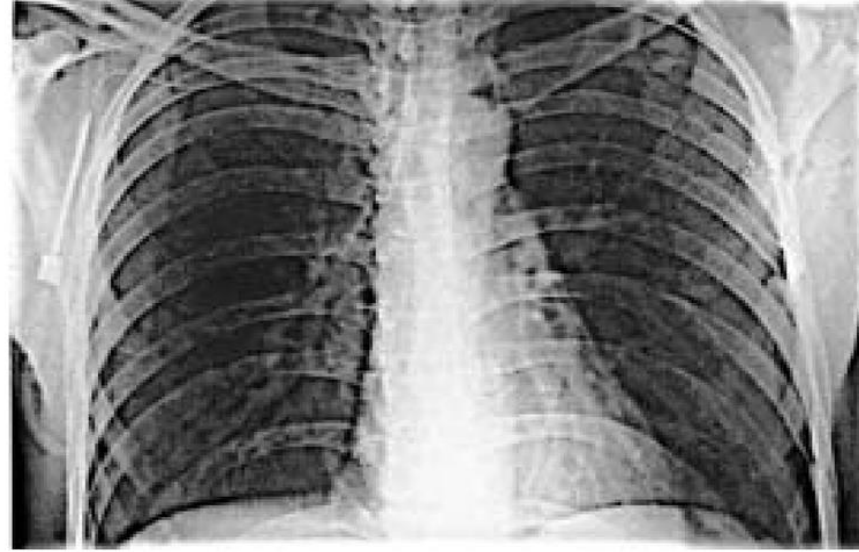
Highpass Filtering Example

Images taken from Gonzalez & Woods, Digital Image Processing (2002)

High frequency
emphasis result



Original image

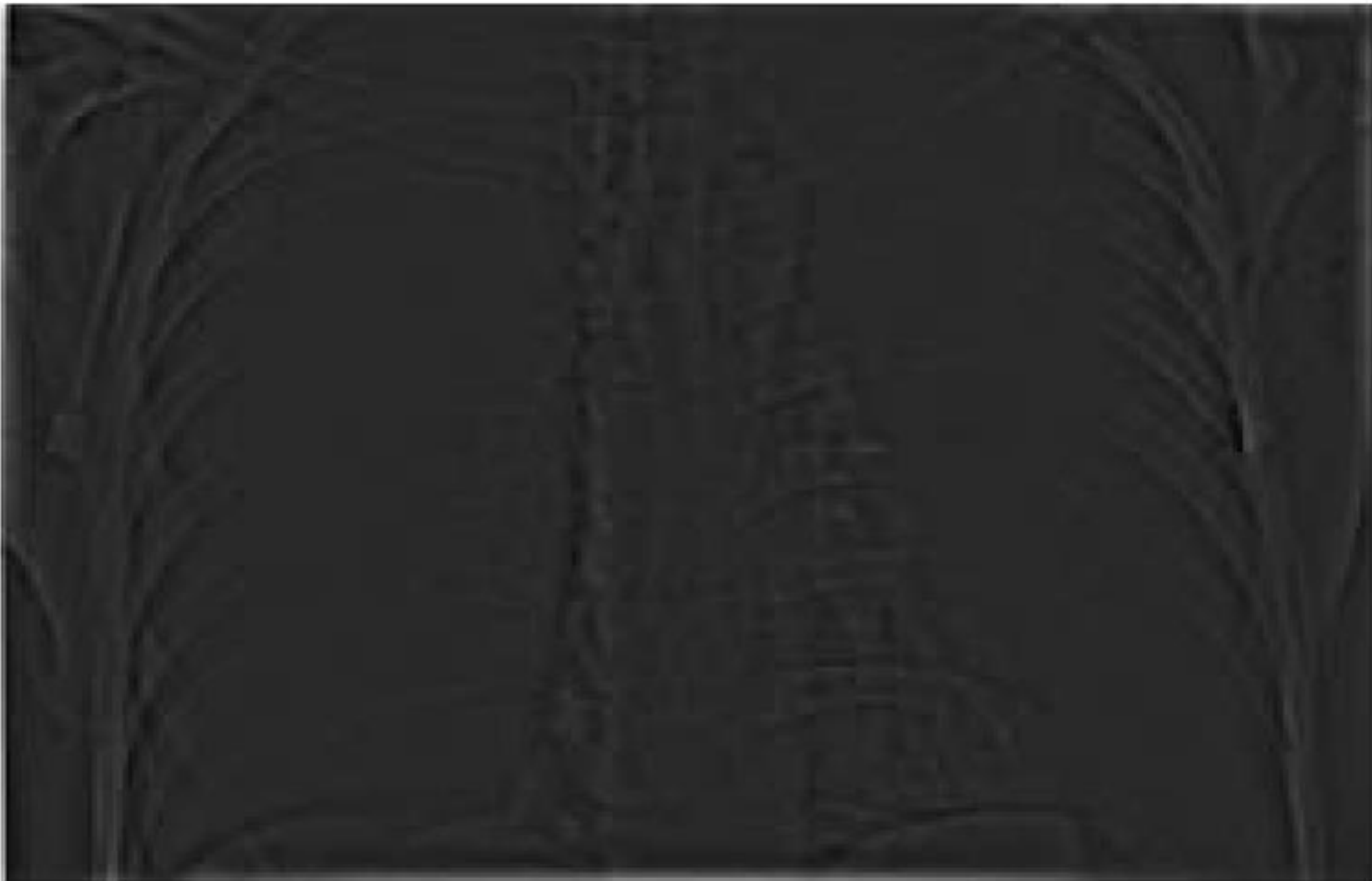


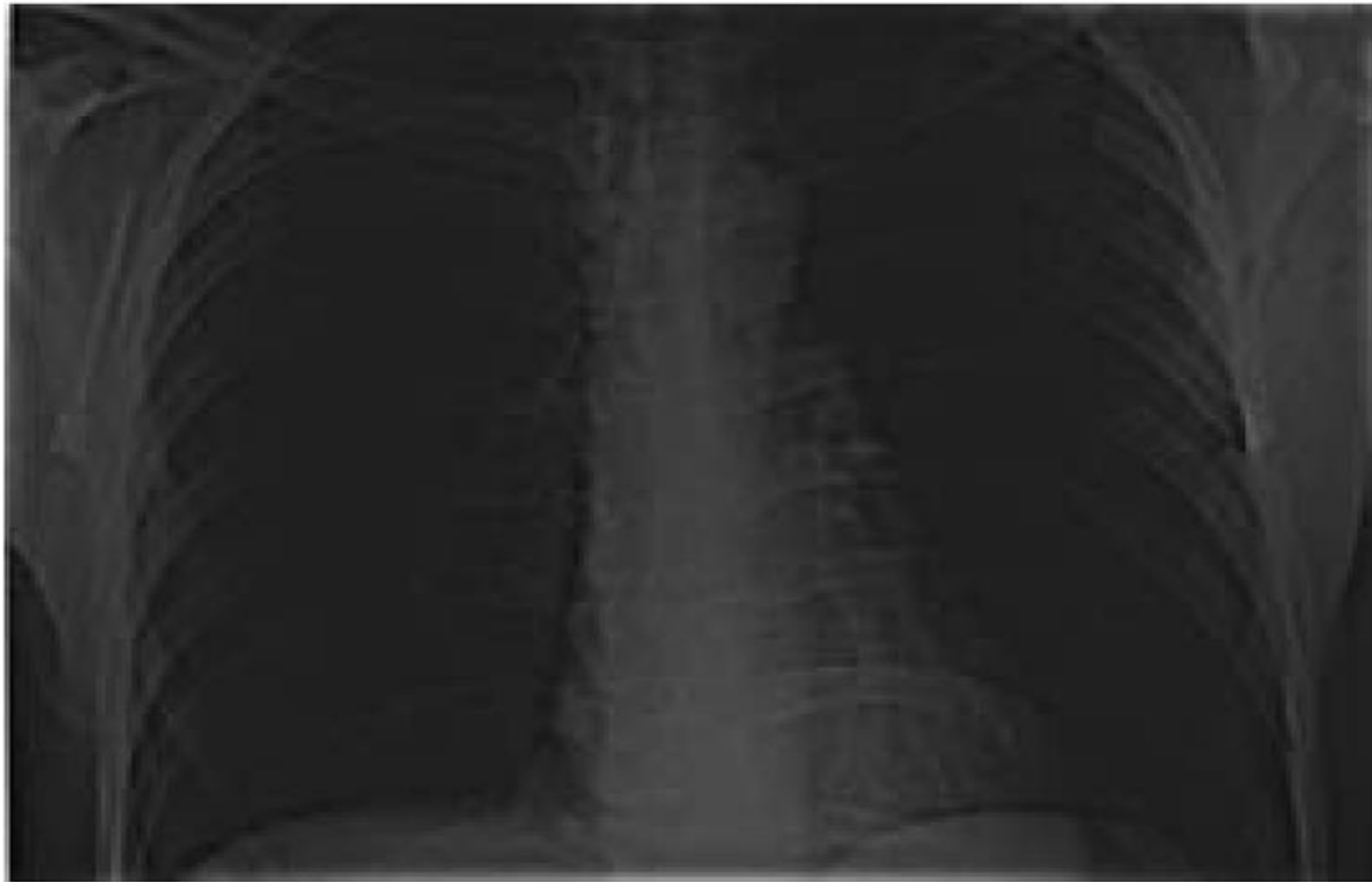
After histogram
equalisation

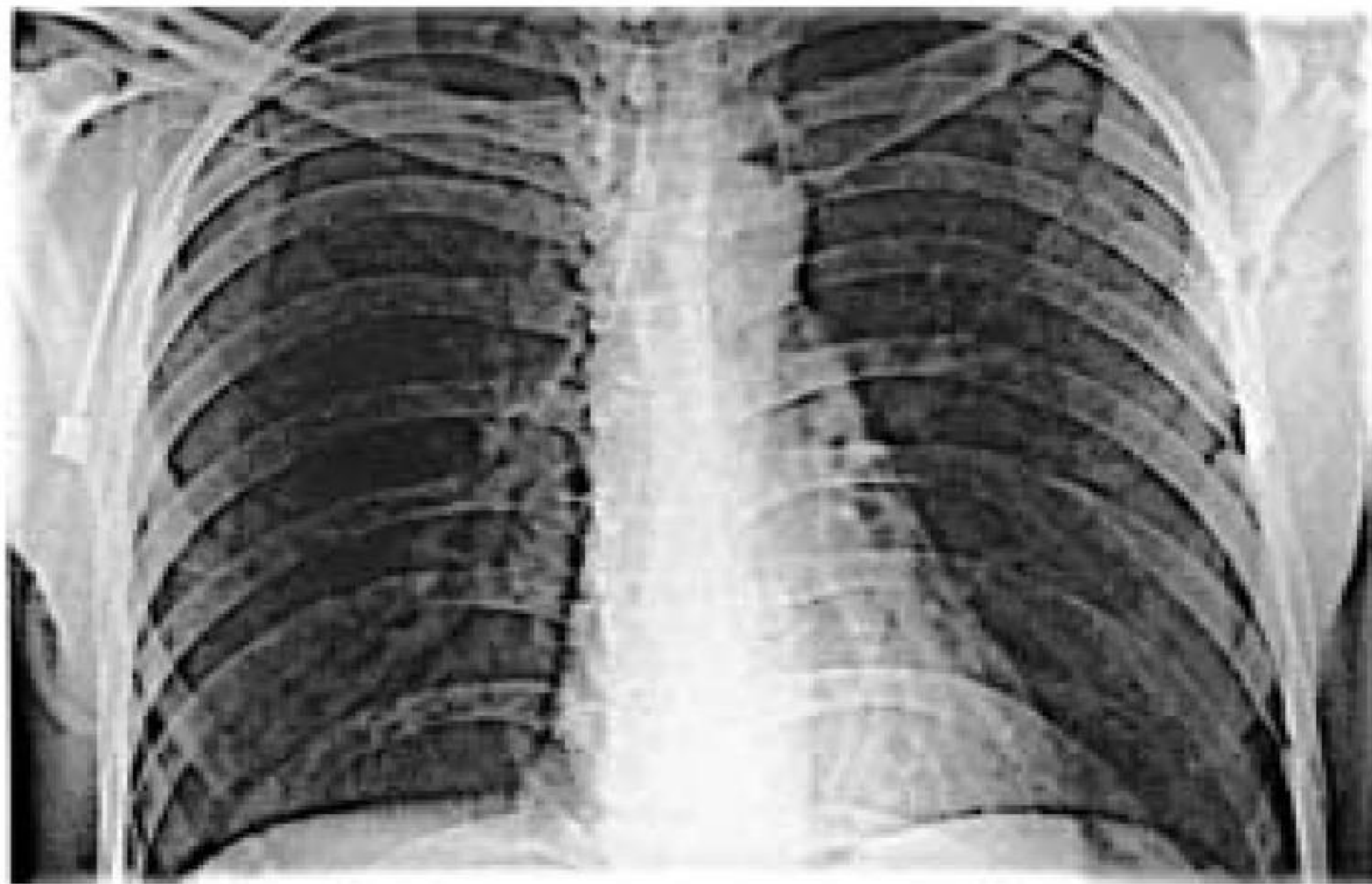


Highpass filtering result



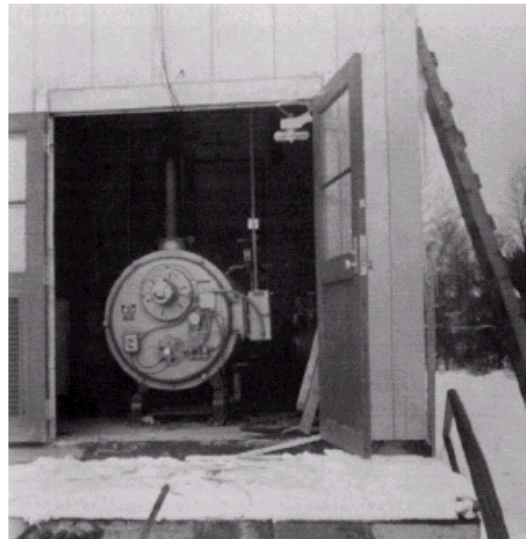






Homomorphic filtering

- Many times, we want to remove shading effects from an image (i.e., due to uneven illumination)
 - Enhance high frequencies
 - Attenuate low frequencies but preserve fine detail.



Homomorphic Filtering (cont'd)

- Consider the following model of image formation:

$$f(x, y) = i(x, y) r(x, y)$$

$i(x, y)$: illumination
 $r(x, y)$: reflection

- In general, the illumination component $i(x, y)$ varies **slowly** and affects **low** frequencies mostly.
- In general, the reflection component $r(x, y)$ varies **faster** and affects **high** frequencies mostly.

IDEA: separate low frequencies due to $i(x, y)$
from high frequencies due to $r(x, y)$

How are frequencies mixed together?

- Low and high frequencies from $i(\mathbf{x}, \mathbf{y})$ and $r(\mathbf{x}, \mathbf{y})$ are mixed together.

$$f(x, y) = i(x, y) r(x, y) \quad \Rightarrow \quad F(u, v) = I(u, v) * R(u, v)$$

- When applying filtering, it is difficult to handle low/high frequencies separately.

$$F(u, v)H(u, v) = [I(u, v) * R(u, v)]H(u, v)$$

Can we separate them?

- Idea:

Take the $\ln()$ of $f(x, y) = i(x, y) r(x, y)$

$$\ln(f(x, y)) = \ln(i(x, y)) + \ln(r(x, y))$$

Steps of Homomorphic Filtering

(1) Take $\ln(f(x, y)) = \ln(i(x, y)) + \ln(r(x, y))$

(2) Apply FT: $F(\ln(f(x, y))) = F(\ln(i(x, y))) + F(\ln(r(x, y)))$

or $Z(u, v) = \text{Illum}(u, v) + \text{Refl}(u, v)$

(3) Apply $H(u, v)$

$$Z(u, v)H(u, v) = \text{Illum}(u, v)H(u, v) + \text{Refl}(u, v)H(u, v)$$

Steps of Homomorphic Filtering (cont'd)

(4) Take Inverse FT:

$$F^{-1}(Z(u, v)H(u, v)) = F^{-1}(\text{Illum}(u, v)H(u, v)) + F^{-1}(\text{Refl}(u, v)H(u, v))$$

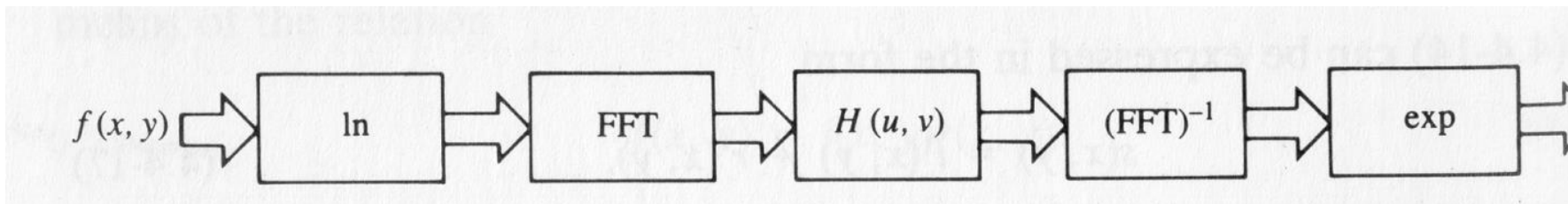
or

$$s(x, y) = i'(x, y) + r'(x, y)$$

(5) Take exp()

$$e^{s(x, y)} = e^{i'(x, y)} e^{r'(x, y)}$$

$$g(x, y) = i_0(x, y) r_0(x, y)$$



Example using high-frequency emphasis

$$H(u, v) = (\gamma_H - \gamma_L) \left[1 - e^{-c[(u^2 + v^2)/D_0^2]} \right] + \gamma_L$$

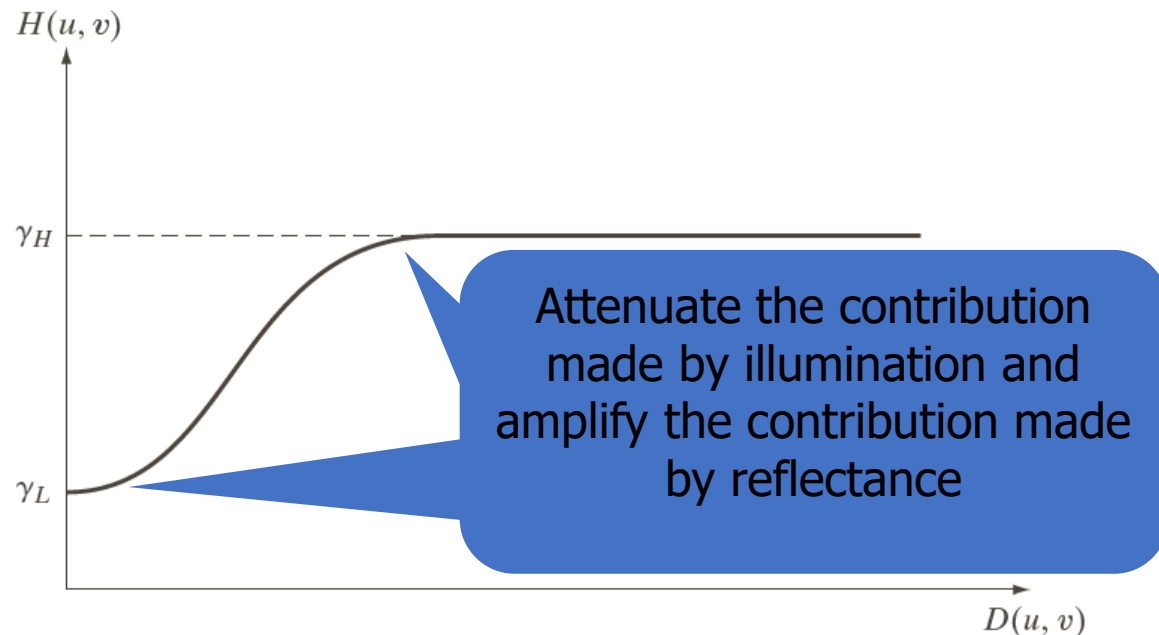


FIGURE 4.61
Radial cross section of a circularly symmetric homomorphic filter function. The vertical axis is at the center of the frequency rectangle and $D(u, v)$ is the distance from the center.

Example 1 homomorphic filters

$$T = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

```
gamma_l = 0.5;  
gamma_h = 1.3;  
cutoff = 50;
```

→ Parameters

```
f1 = log(1+img);
```

```
0.6931 0 0.6931 0  
0.6931 0 0.6931 0  
0.6931 0 0.6931 0  
0.6931 0 0.6931 0
```

Example 1 homomorphic filters

- After the transform center arrangement we got the results as follows.

$$\begin{bmatrix} 0.6931 & 0 & 0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 \\ -1 & 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 0.6931 & 0 & 0.6931 & 0 \\ -0.6931 & 0 & -0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \\ -0.6931 & 0 & -0.6931 & 0 \end{bmatrix}$$

Example 1 homomorphic filters

- Compute the DFT of the image.
- $F(u, v) = \text{kernel} * f(x, y) * \text{kernel}^T$

$$f'(u, v) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -j \\ 1 & j & -1 & -j \end{bmatrix} \times \begin{bmatrix} 0.6931 & 0 & 0.6931 & 0 \\ -0.6931 & 0 & -0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \\ -0.6931 & 0 & -0.6931 & 0 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -j \\ 1 & j & -1 & -j \end{bmatrix}$$

Example 1 homomorphic filters

$$f(u,v) = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix} \times \begin{pmatrix} 0.6931 & 0 & 0.6931 & 0 \\ 0.6931 & 0 & -0.6931 & 0 \\ 0.6931 & 0 & 0.6931 & 0 \\ 0.6931 & 0 & -0.6931 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix}$$

$$F(u,v) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 5.5452 & 0 & 5.5452 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

Distance measure

$$U = \begin{bmatrix} -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \\ -2 & -1 & 0 & 1 \end{bmatrix}$$

$$V = \begin{bmatrix} -2 & -2 & -2 & -2 \\ -1 & -1 & -1 & -1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$$D(U, V) = \begin{bmatrix} (-2, -2) & (-1, -2) & (0, -2) & (1, -2) \\ (-2, -1) & (-1, -1) & (0, -1) & (1, -1) \\ (-2, 0) & (-1, 0) & (0, 0) & (1, 0) \\ (-2, 1) & (-1, 1) & (0, 1) & (1, 1) \end{bmatrix}$$

Distance measure

- Now compute the distances
- $D(u, v) = \sqrt{u.^2 + v.^2}$
- $D(-2, -2) = (-2)^2 + (-2)^2 = (8)^{1/2} = 2.82$
- $D(-1, -2) = (-1)^2 + (-2)^2 = (5)^{1/2} = 2.23$
- $D(0, -2) = (0)^2 + (-2)^2 = (4)^{1/2} = 2$
- $D(1, -2) = (1)^2 + (-2)^2 = 2.23$
- $D(-2, -1) = (-2)^2 + (-1)^2 = 1.41$
- $D(-1, -1) = (-1)^2 + (-1)^2 = 1.41$

Distance measure

- $D(0, -1) = (0)^2 + (-1)^2 = 1$
- $D(1, -1) = (1)^2 + (-1)^2 = 1.41$
- $D(-2, 0) = (-2)^2 + (0)^2 = 2$
- $D(-1, 0) = (-1)^2 + (0)^2 = 1$
- $D(0, 0) = (0)^2 + (0)^2 = 0$
- $D(1, 0) = (1)^2 + (0)^2 = 1$
- $D(-2, 1) = (-2)^2 + (1)^2 = 2.23$
- $D(-1, 1) = (-1)^2 + (1)^2 = 1.41$
- $D(0, 1) = (0)^2 + (1)^2 = 1$
- $D(1, 1) = (1)^2 + (1)^2 = 1.41$

$$H(u, v) = (\gamma_H - \gamma_L) \left[1 - e^{-c \left[(u^2 + v^2) / D_0^2 \right]} \right] + \gamma_L$$

$$D(u, v) = \begin{bmatrix} 2.8284 & 2.2361 & 2 & 2.2361 \\ 2.2361 & 1.4142 & 1 & 1.4142 \\ 2 & 1 & 0 & 1 \\ 2.2361 & 1.4142 & 1 & 1.4142 \end{bmatrix} = \begin{bmatrix} (-2, -2) & (-1, -2) & (0, -2) & (1, -2) \\ (-2, -1) & (-1, -1) & (0, -1) & (1, -1) \\ (-2, 0) & (-1, 0) & (0, 0) & (1, 0) \\ (-2, 1) & (-1, 1) & (0, 1) & (1, 1) \end{bmatrix}$$

$$H = (\text{gamma_h} - \text{gamma_l}) .* (1 - \exp(-c .* (D_{uv}.^2 / (D0^2)))) + \text{gamma_l};$$

HMF

$$h(u, v) = \begin{bmatrix} 0.5026 & 0.5016 & 0.5013 & 0.5016 \\ 0.5016 & 0.5006 & 0.5003 & 0.5006 \\ 0.5013 & 0.5003 & 0.5000 & 0.5003 \\ 0.5016 & 0.5006 & 0.5003 & 0.5006 \end{bmatrix}$$

Example 1 homomorphic filters

$$g(u, v) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 5.5452 & 0 & 5.5452 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} * \begin{bmatrix} 0.5026 & 0.5016 & 0.5013 & 0.5016 \\ 0.5016 & 0.5006 & 0.5003 & 0.5006 \\ 0.5013 & 0.5003 & 0.5000 & 0.5003 \\ 0.5016 & 0.5006 & 0.5003 & 0.5006 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 2.7797 & 0 & 2.7726 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Example 1 homomorphic filters

$$\text{DFT}^{-1} = \begin{bmatrix} 5.5523 & 0.0071 & 5.5523 & 0.0071 \\ -5.5523 & -0.0071 & -5.5523 & -0.0071 \\ 5.5523 & 0.0071 & 5.5523 & 0.0071 \\ -5.5523 & -0.0071 & -5.5523 & -0.0071 \end{bmatrix}$$

```
g1 = exp(DFT_Inverse) - 1;
```

$$g_1 = \begin{bmatrix} 256.8220 & 0.0071 & 256.8220 & 0.0071 \\ -0.9961 & -0.0071 & -0.9961 & -0.0071 \\ 256.8220 & 0.0071 & 256.8220 & 0.0071 \\ -0.9961 & -0.0071 & -0.9961 & -0.0071 \end{bmatrix} \Rightarrow$$

```
g2 = g1 ./ max(g1(:));
```

$$\begin{bmatrix} 1.0000 & 0.0000 & 1.0000 & 0.0000 \\ -0.0039 & -0.0000 & -0.0039 & -0.0000 \\ 1.0000 & 0.0000 & 1.0000 & 0.0000 \\ -0.0039 & -0.0000 & -0.0039 & -0.0000 \end{bmatrix}$$

Resultant HNF image

Homomorphic Filtering

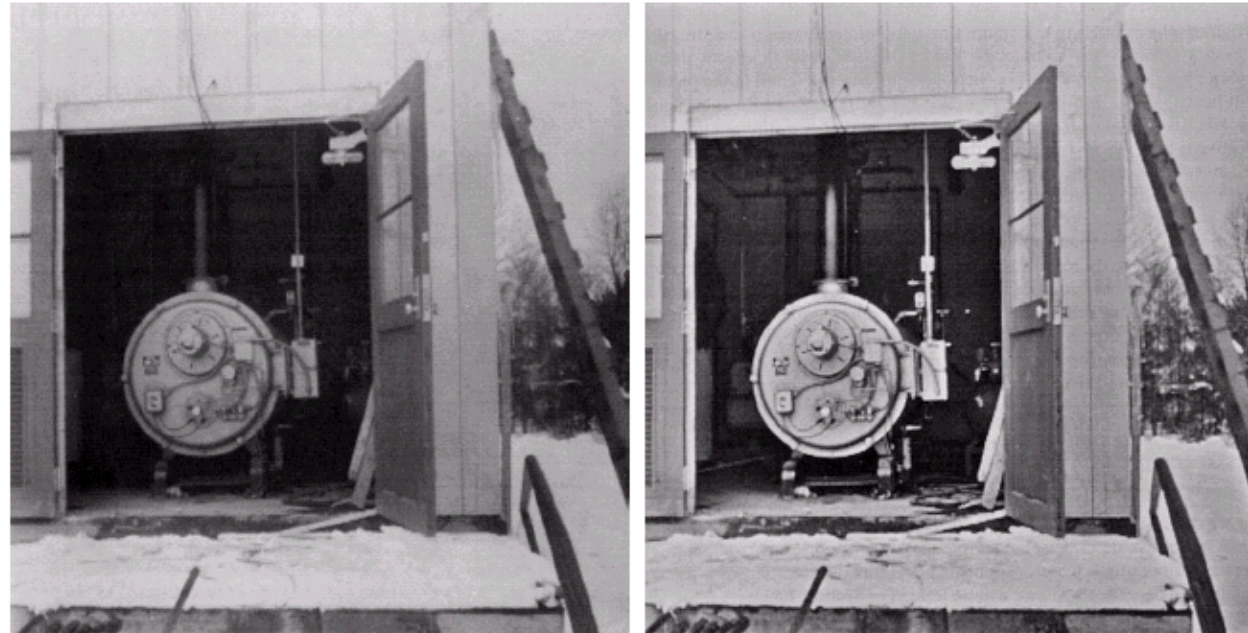
- In homomorphic filtering, γ_h and γ_l are parameters used to control the high-frequency and low-frequency emphasis of the filter.
- γ_h , also known as the high-frequency gamma, is a parameter that controls the emphasis given to the high-frequency components of the image. Increasing γ_h results in a stronger emphasis on the high-frequency components, making the image appear sharper and more detailed.
- γ_l , also known as the low-frequency gamma, is a parameter that controls the emphasis given to the low-frequency components of the image. Increasing γ_l results in a stronger emphasis on the low-frequency components, making the image appear smoother and less detailed.
- The choice of γ_h and γ_l values depends on the specific application and the characteristics of the image being processed. In general, higher values of γ_h and lower values of γ_l are used to enhance the edges and details in the image, while lower values of γ_h and higher values of γ_l are used to smooth out the image and reduce noise.

Homomorphic Filtering: Example

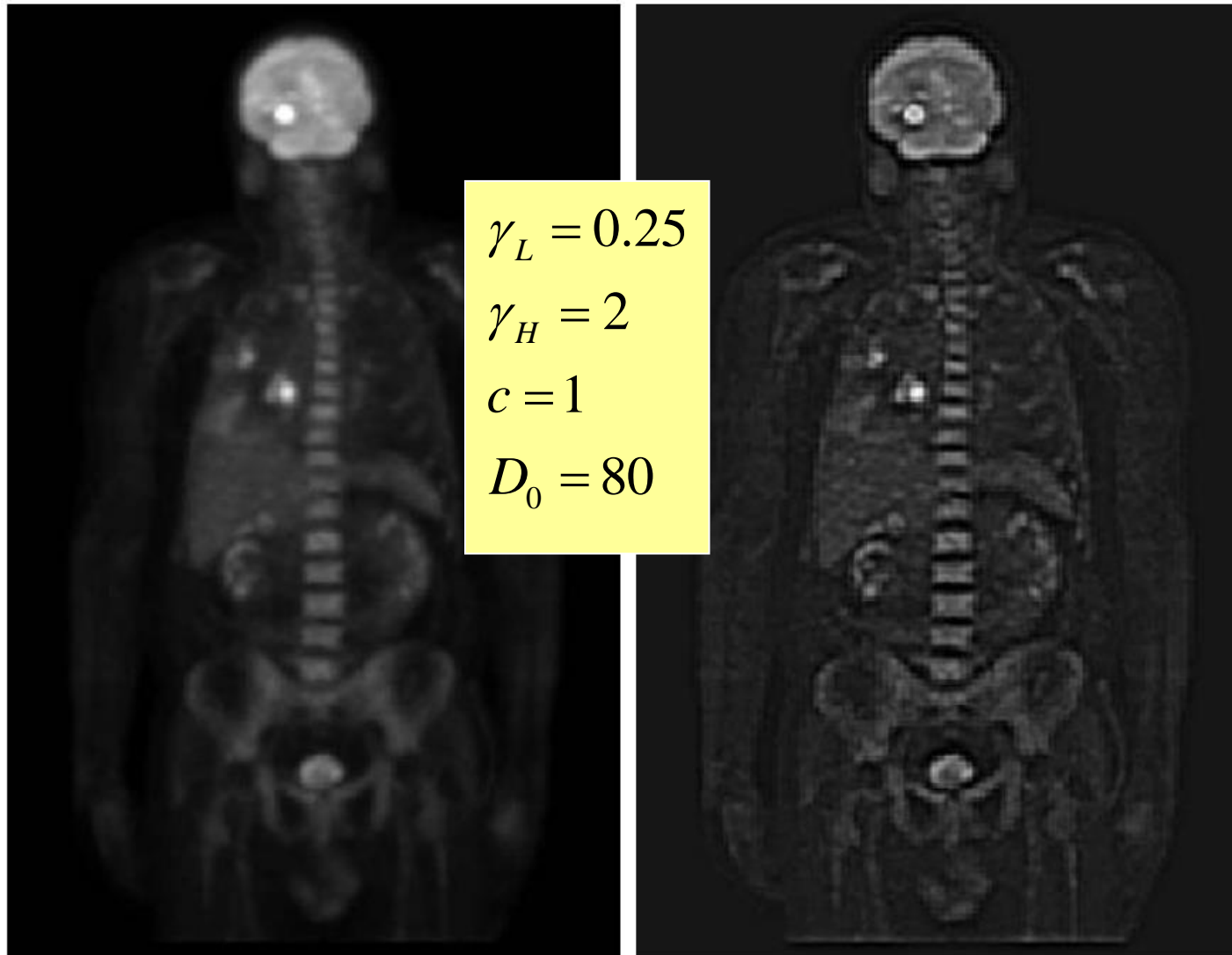
a b

FIGURE 4.33

(a) Original image. (b) Image processed by homomorphic filtering (note details inside shelter). (Stockham.)



Homomorphic Filtering: Example



a b

FIGURE 4.62

(a) Full body PET scan. (b) Image enhanced using homomorphic filtering. (Original image courtesy of Dr. Michael E. Casey, CTI PET Systems.)

Summery of the lecture

- Frequency domain Filters
- Ideal Lowpass and High pass Filters
- Butterworth Lowpass and High pass Filters
- Gaussian Lowpass and High pass Filters
- Homomorphic Filtering