

EE 4211 Computer Vision

Lecture 3B: Image enhancement (Frequency)

Semester B, 2021-2022

Spatial Domain vs. Frequency Domain

- **Spatial Domain** (image plane)
 - Techniques are based on direct **manipulation of pixels** in an image
- **Frequency Domain**
 - Techniques are based on modifying the **spectral transform** (in our course, we'll use Fourier transform) of an image
- There are some enhancement techniques based on various combinations of methods from these 2 domains

Lecture Outline

- Image Enhancement in Frequency Domain
- Filtering in Frequency Domain
 - Low-Pass Filtering
 - High-Pass Filtering
 - Laplacian Filtering
 - Homomorphic Filtering
 - Selective Filtering – Bandpass/Bandreject, Notch Filters

Image Enhancement in Frequency Domain

- Images after transformation to frequency domain can be modified with frequency filters
- Nature of periodicity & conjugate symmetry.
- So, origin of spectrum is always shifted for
 - Display purpose
 - Filtering purpose

Image Enhancement in Frequency Domain

- To filter an image in the frequency domain:
 - Compute $F(u,v)$ the DFT of the image
 - Multiply $F(u,v)$ by a filter function $H(u,v)$
 - Compute the inverse DFT of the result

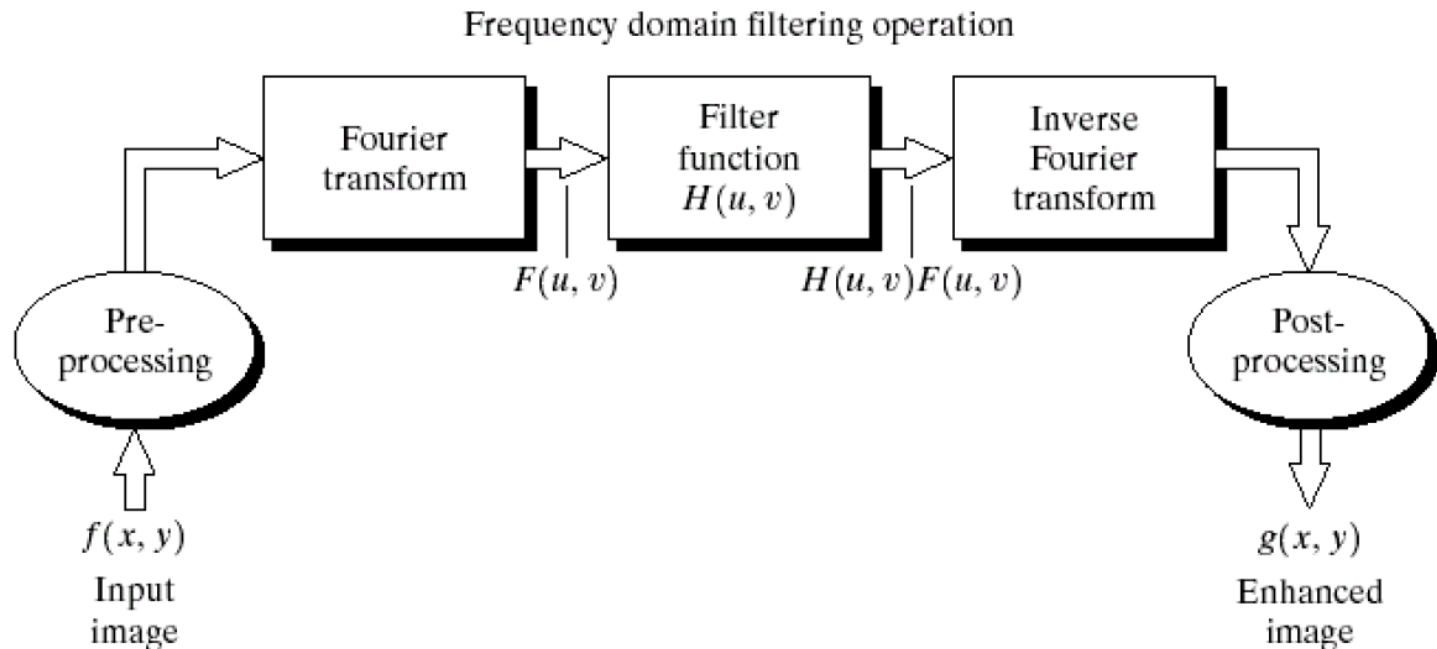


Image Enhancement in Frequency Domain

- Steps taken:
- Shift the origin of spectrum by multiplying image $f(x,y)$ by $(-1)^{x+y}$ before performing transformation to frequency domain.

$$f(x,y)(-1)^{(x+y)} \quad F\left(u - \frac{M}{2}, v - \frac{N}{2}\right)$$

Image Enhancement in Frequency Domain

- $G(u,v)$, the frequency spectrum obtained after applying frequency filter $H(u,v)$ to frequency-transform image $F(u,v)$ by multiplication, is defined as:

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

- The filtered image, $g(x,y)$ can then be obtained by performing the inverse transform to $G(u,v)$:

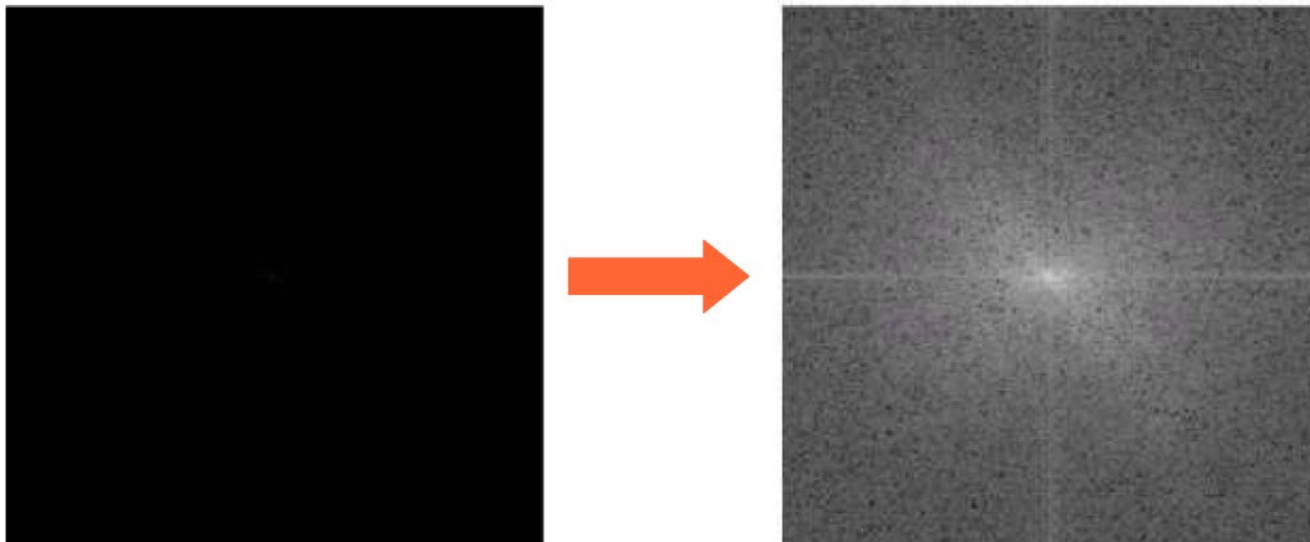
$$g(x,y) = G^{-1}(u,v)$$

- Any shifting to origin performed before filtering should also be reversed after filtering.

Image Enhancement in Frequency Domain

- Steps taken:
- Enhance the visual information of the transformed image $G(u,v)$ using log transform:

$$D(u, v) = k \log[1 + |G(u, v)|]$$



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Low-Pass Filter (in Freq. Domain)

- Lowpass filter – remove high-frequency information, or allow low-frequency information to PASS through
- Useful for removing noise in images
- Also have undesired effect of blurring an image

Ideal Low-Pass Filter (ILPF)

- An ideal low-pass filter contain only 1's and 0's – 1 for lower frequency and 0 for high frequency
- 2-D ideal lowpass filter (ILPF) is defined as:

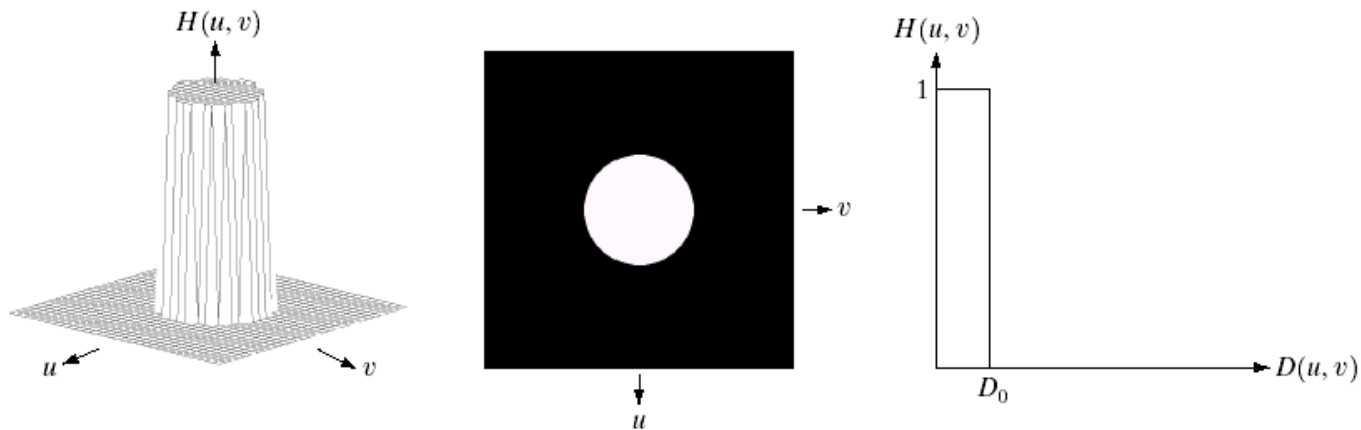
$$H(u, v) = \begin{cases} 1 & \text{if } D(u, v) \leq D_0 \\ 0 & \text{if } D(u, v) > D_0 \end{cases}$$

- where D_0 is a positive constant and $D(u,v)$ is the distance from point (u,v) to the origin (center) of the frequency rectangle. It is denoted as

$$D(u, v) = \sqrt{(u - M / 2)^2 + (v - N / 2)^2}$$

Ideal Low-Pass Filter (ILPF)

- Ideal Low pass filter (ILPF): all frequencies inside a circle of radius D_0 are passed with no attenuation



a b c

- (a) Perspective plot of an ideal low pass filter transfer function
- (b) Filter displayed as an image
- (c) Filter radial cross section

Images Filtered by ILPFs

Original

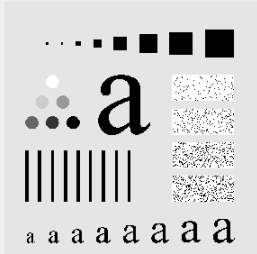


Image after f=20



Image after f=40

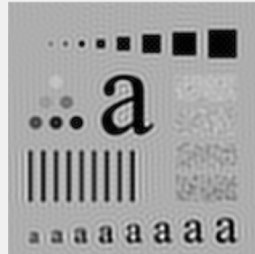
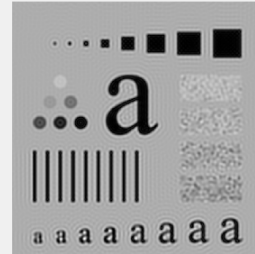
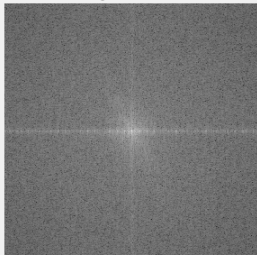


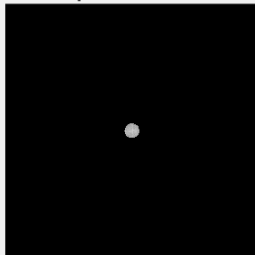
Image after f=60



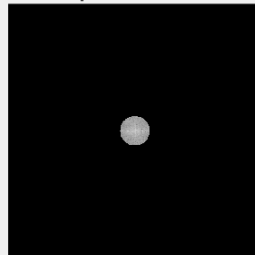
Spectrum



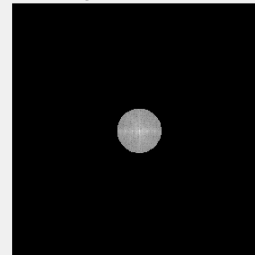
Spectrum f=20



Spectrum f=40



Spectrum f=60



```
close all;
I=imread('test1.tif');
ff=imidealflpf(I,20);
out=imfreqfilt(I,ff);
ff1=imidealflpf(I,40);
out1=imfreqfilt(I,ff1);
ff2=imidealflpf(I,60);
out2=imfreqfilt(I,ff2);

figure;
subplot(2,4,1);imshow(mat2gray(abs(I)));title('Original ');
subplot(2,4,2);imshow(mat2gray(abs(out)));title('Image after f=20');
subplot(2,4,3);imshow(mat2gray(abs(out1)));title('Image after f=40');
subplot(2,4,4);imshow(mat2gray(abs(out2)));title('Image after f=60');

%show spectrum (fft--fftshift--abs--log)
If=fft2(I);
If=fftshift(If);
If=abs(If);
If=log(1+If);

outf=fft2(out);
outf=fftshift(outf);
outf=abs(outf);
outf=log(1+outf);

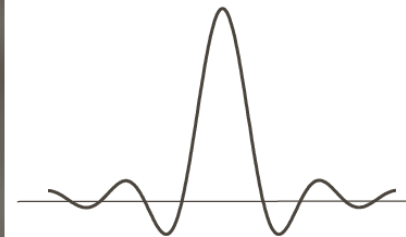
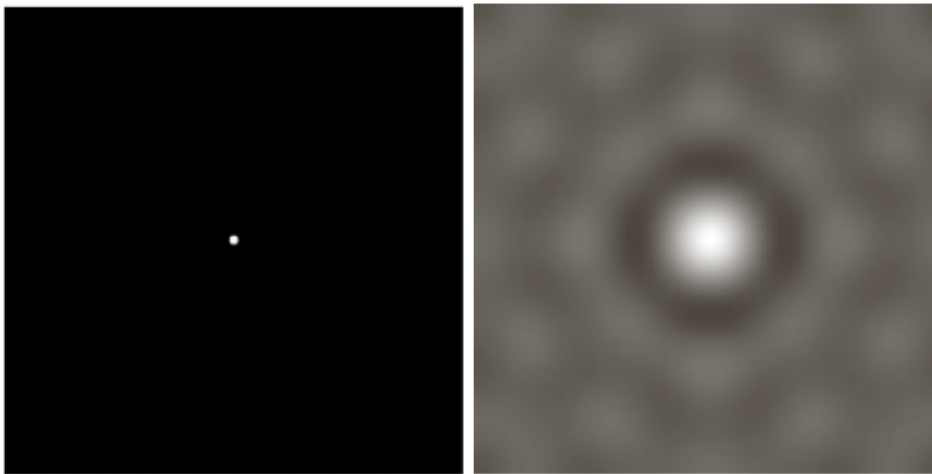
out1f=fft2(out1);
out1f=fftshift(out1f);
out1f=abs(out1f);
out1f=log(1+out1f);

out2f=fft2(out2);
out2f=fftshift(out2f);
out2f=abs(out2f);
out2f=log(1+out2f);

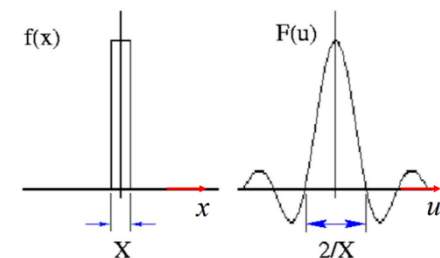
subplot(2,4,5);imshow(If,[]);title('Spectrum');
```

Ideal Low-Pass Filter (ILPF)

- An ideal filter has undesired artifacts in images
- Presence of ripples/waves whenever there are boundaries in the image – “ringing effect”



Reminder



Butterworth Low-Pass Filter (BLPF)

- Transfer function of a Butterworth lowpass filter (BLPF) of order n , with D_0 cutoff frequency, is defined as

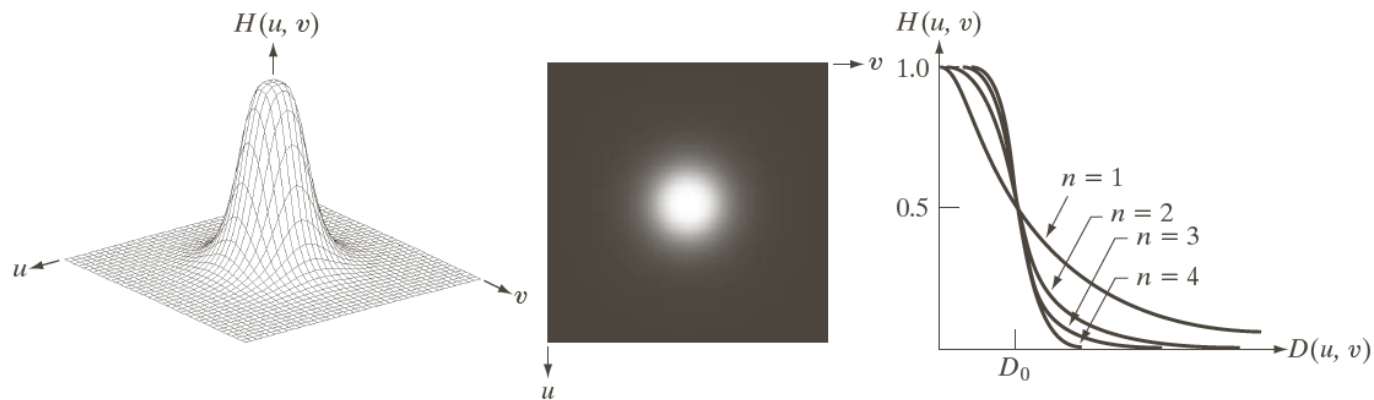
$$H(u, v) = \frac{1}{1 + [D(u, v) / D_o]^{2n}}$$

- where $D(u,v)$ is the distance from origin of spectrum
- **Butterworth lowpass filter** – Can specify **order of filter**, which determines steepness of slope in the transition of the filter function
 - Higher order of filter – steeper slope – closer to ideal filter

Butterworth Low-Pass Filter (BLPF)

■ Advantages

- Reduces “ringing” while keeping clear cutoff
- Tradeoff between amount of ringing and sharpness of cutoff



a b c

- (a) Perspective plot of a Butterworth low pass filter transfer function
(b) Filter displayed as an image
(c) Filter radial cross section of orders 1 through 4

Butterworth Low-Pass Filter (BLPF)

Original

Image after btw D0=10

Image after btw D1=20

Image after btw D2=30

```
%% btw low pass
I=imread('test1.tif');
D1=10;
D2=20;
D3=30;
ff=imbtwflpf(I,D1);
out=imfreqfilt(I,ff);
ff1=imbtwflpf(I,D2);
out1=imfreqfilt(I,ff1);
ff2=imbtwflpf(I,D3);
out2=imfreqfilt(I,ff2);
figure;
subplot(2,4,1);imshow(I,[]);title('Original ');
subplot(2,4,2);imshow(mat2gray(abs(out)));title('Image after btw D0=10');
subplot(2,4,3);imshow(mat2gray(abs(out1)));title('Image after btw D1=20');
subplot(2,4,4);imshow(mat2gray(abs(out2)));title('Image after btw D2=30');

%show specturm (fft--fftshift--abs--log)
If=fft2(I);
If=fftshift(If);
If=abs(If);
If=log(1+If);

outf=fft2(out);
outf=fftshift(outf);
outf=abs(outf);
outf=log(1+outf);

out1f=fft2(out1);
out1f=fftshift(out1f);
out1f=abs(out1f);
out1f=log(1+out1f);

out2f=fft2(out2);
out2f=fftshift(out2f);
out2f=abs(out2f);
out2f=log(1+out2f);

subplot(2,4,5);imshow(If,[]);title('Spectrum');
```

Spectrum

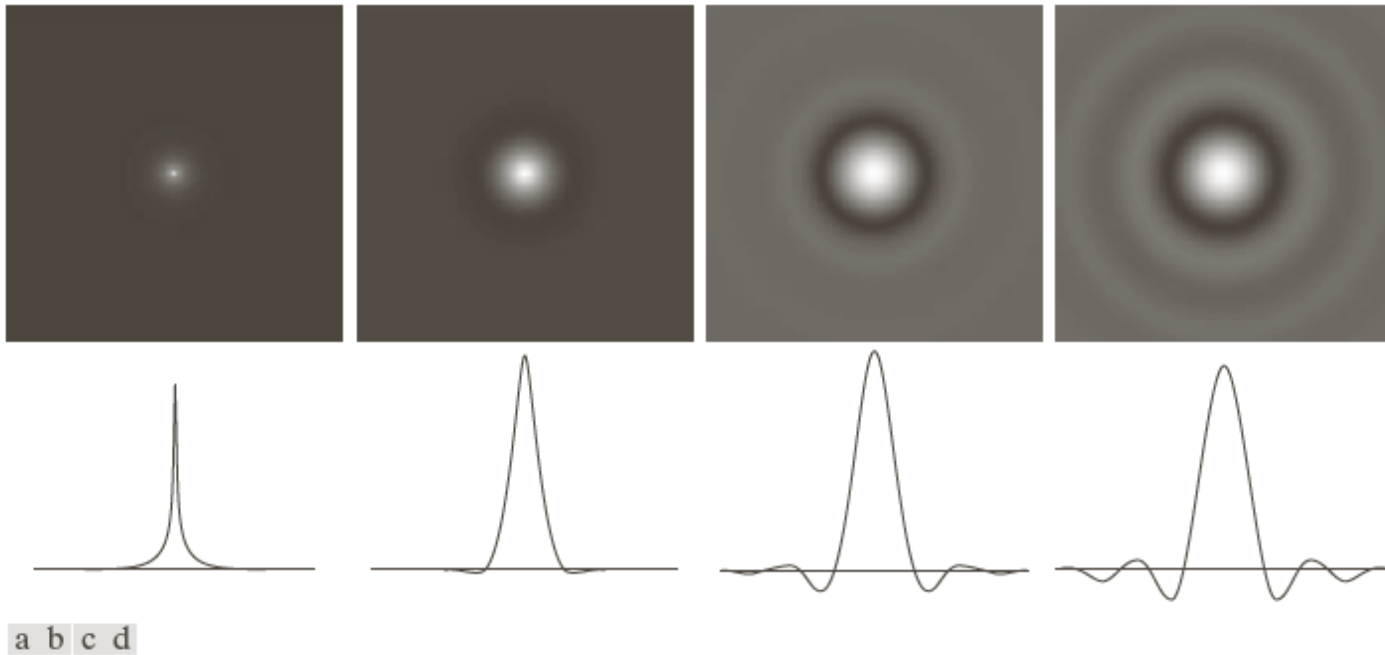
Spectrum btw D0=10

Spectrum btw D1=20

Spectrum btw D2=30

Butterworth Low-Pass Filter (BLPF)

- Ringing properties increase if we increase the BLPF filter order, n



(a)-(d) Spatial representation of BLPFS of order 1, 2, 5, 20 and corresponding intensity profiles

Gaussian Low-Pass Filter (GLPF)

- **Gaussian lowpass filter** in 2-D is defined as

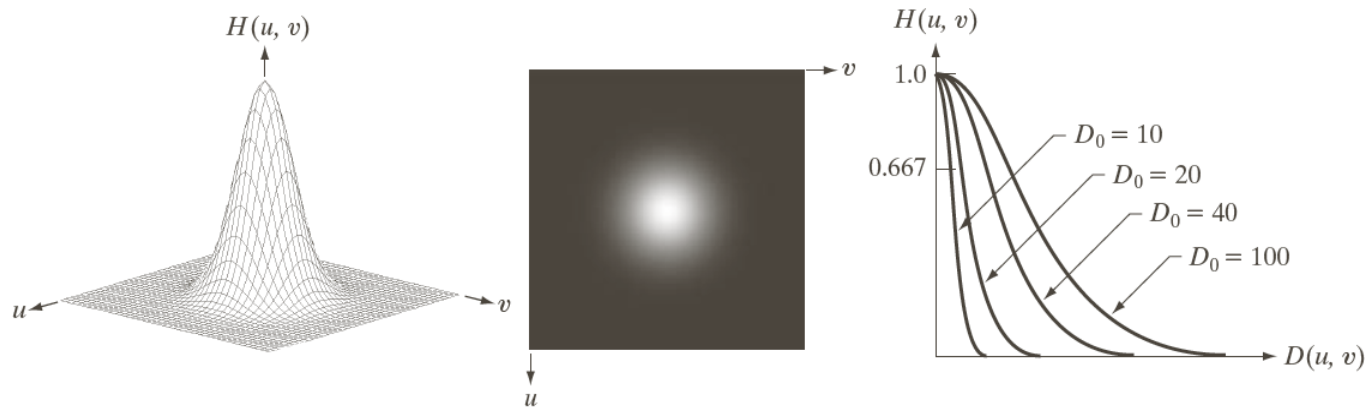
$$H(u, v) = e^{-D^2(u, v) / 2\sigma^2}$$

- where $D(u, v)$ is the distance from the origin of spectrum, σ is the measure of spread of the Gaussian curve
- By letting $\sigma = D_0$, where D_0 is the cutoff frequency, we get

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

Gaussian Low-Pass Filter (GLPF)

- A Gaussian in the spatial domain also has the form of a Gaussian in the frequency domain
- No ringing, but allows high frequencies to pass



a b c

- (a) Perspective plot of a GLPF transfer function
- (b) Filter displayed as an image
- (c) Filter radial cross sections for various values of D_0

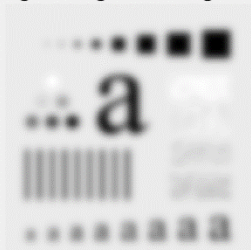
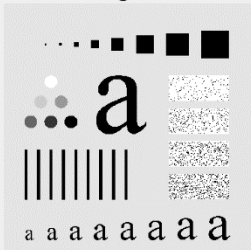
Images Filtered by GLPFs

Original

Image after gaussian sigma=10

Image after gaussian sigma=20

Image after gaussian sigma=30



```
%% ideal gauss low pass filter
I=imread('test1.tif');
sigma1=10;
sigma2=20;
sigma3=30;
ff=imgaussfilt(I,sigma1);
out=imfreqfilt(I,ff);
ff1=imgaussfilt(I,sigma2);
out1=imfreqfilt(I,ff1);
ff2=imgaussfilt(I,sigma3);
out2=imfreqfilt(I,ff2);
figure;
subplot(2,4,1);imshow(I,[]);title('Original ');
subplot(2,4,2);imshow(mat2gray(abs(out)));title('Image after gaussian sigma=10');
subplot(2,4,3);imshow(mat2gray(abs(out1)));title('Image after gaussian sigma=20');
subplot(2,4,4);imshow(mat2gray(abs(out2)));title('Image after gaussian sigma=30');

%show spectrum (fft--fftshift--abs--log)
If=fft2(I);
If=fftshift(If);
If=abs(If);
If=log(1+If);

outf=fft2(out);
outf=fftshift(outf);
outf=abs(outf);
outf=log(1+outf);

out1f=fft2(out1);
out1f=fftshift(out1f);
out1f=abs(out1f);
out1f=log(1+out1f);

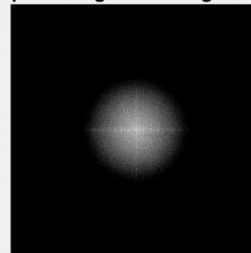
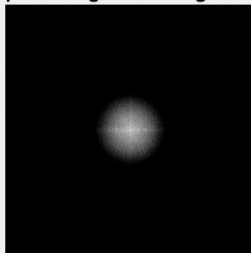
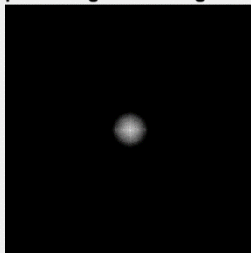
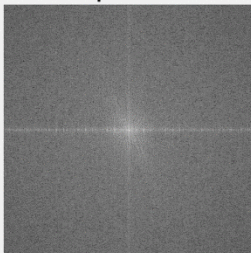
out2f=fft2(out2);
out2f=fftshift(out2f);
out2f=abs(out2f);
out2f=log(1+out2f);
```

Spectrum

Spectrum gaussian sigma=10

Spectrum gaussian sigma=20

Spectrum gaussian sigma=30



Lecture Outline

- Image Enhancement in Frequency Domain
- Filtering in Frequency Domain
 - Low-Pass Filtering
 - High-Pass Filtering
 - Laplacian Filtering
 - Homomorphic Filtering
 - Selective Filtering – Bandpass/Bandreject, Notch Filters

High-Pass Filter (in Freq. Domain)

- Highpass filter – **remove low-frequency information**, or allow **HIGH-frequency** information to **PASS** through
- Useful for sharpening, edge enhancement
- Transfer function of the highpass filters can be obtained using the relation

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

- where $H_{lp}(u, v)$ is the transfer function of the corresponding lowpass filter

High-Pass Filters

- Ideal highpass filter

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

- Butterworth highpass filter of order n and with cutoff frequency at a distance D_0 from the origin:

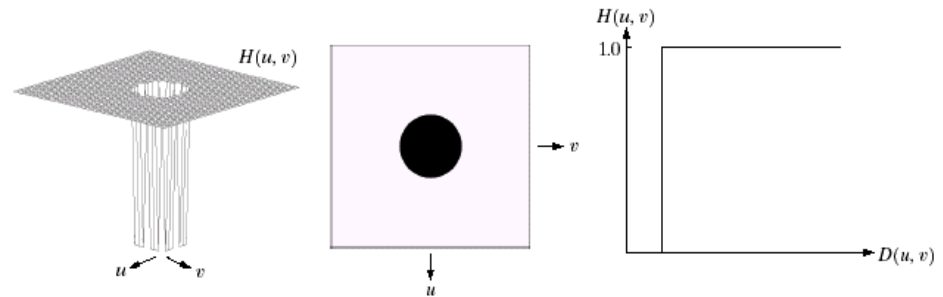
$$H(u, v) = 1 - \frac{1}{1 + [D_0 / D(u, v)]^{2n}}$$

- Gaussian highpass filter with cutoff frequency at a distance D_0 from the origin

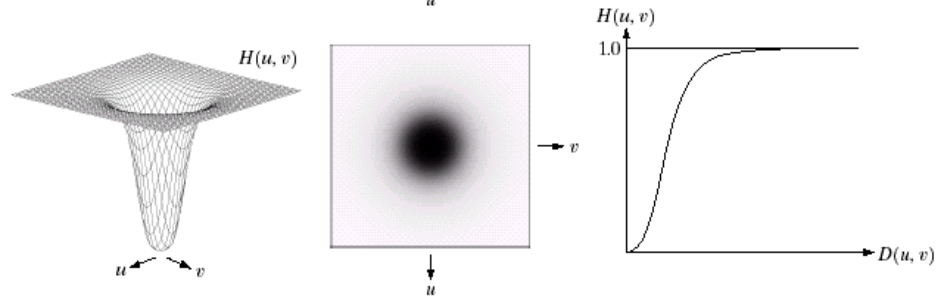
$$H(u, v) = 1 - e^{-D^2(u, v) / 2D_0^2}$$

High-Pass Filters

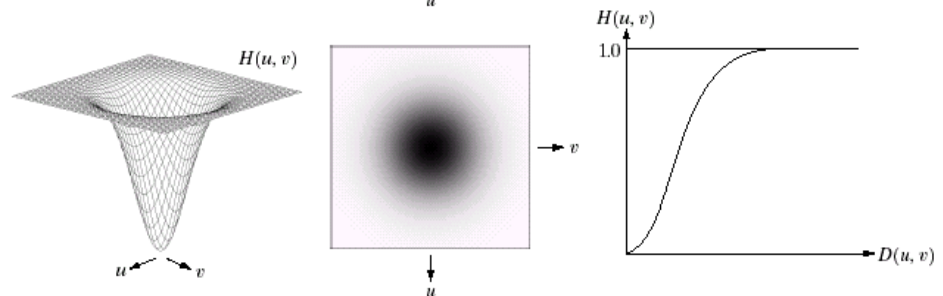
Ideal highpass filter



Butterworth highpass filter



Gaussian highpass filter



Filtering Results by IHPF

Original

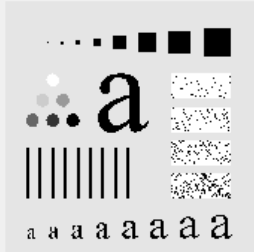


Image after f=20



Image after f=40

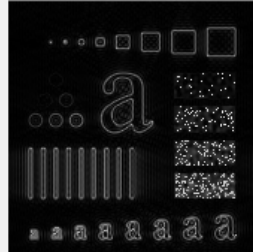
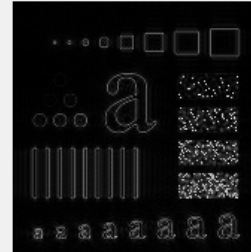
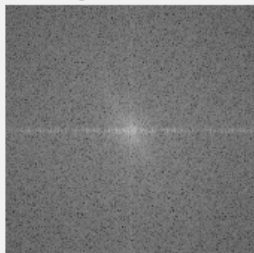


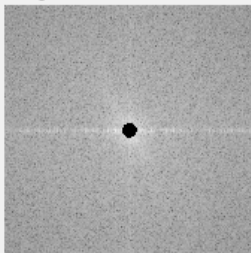
Image after f=60



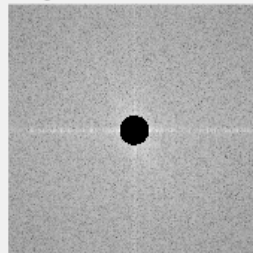
Spectrum



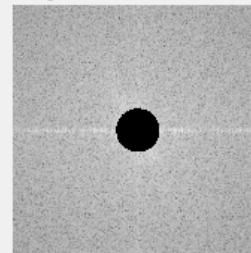
Spectrum f=20



Spectrum f=40



Spectrum f=60



```
%% ideal high filter
I=imread('test1.tif');
ff=imidealfilpf(I,20);ff=1-ff;
out=imfreqfilt(I,ff);
ff1=imidealfilpf(I,40);ff1=1-ff1;
out1=imfreqfilt(I,ff1);
ff2=imidealfilpf(I,60);ff2=1-ff2;
out2=imfreqfilt(I,ff2);

figure;
subplot(2,4,1);imshow(mat2gray(abs(I)));title('Original ');
subplot(2,4,2);imshow(mat2gray(abs(out)));title('Image after f=20');
subplot(2,4,3);imshow(mat2gray(abs(out1)));title('Image after f=40');
subplot(2,4,4);imshow(mat2gray(abs(out2)));title('Image after f=60');

%show spectrum (fft---fftshift---abs---log)
If=fft2(I);
If=fftshift(If);
If=abs(If);
If=log(1+If);

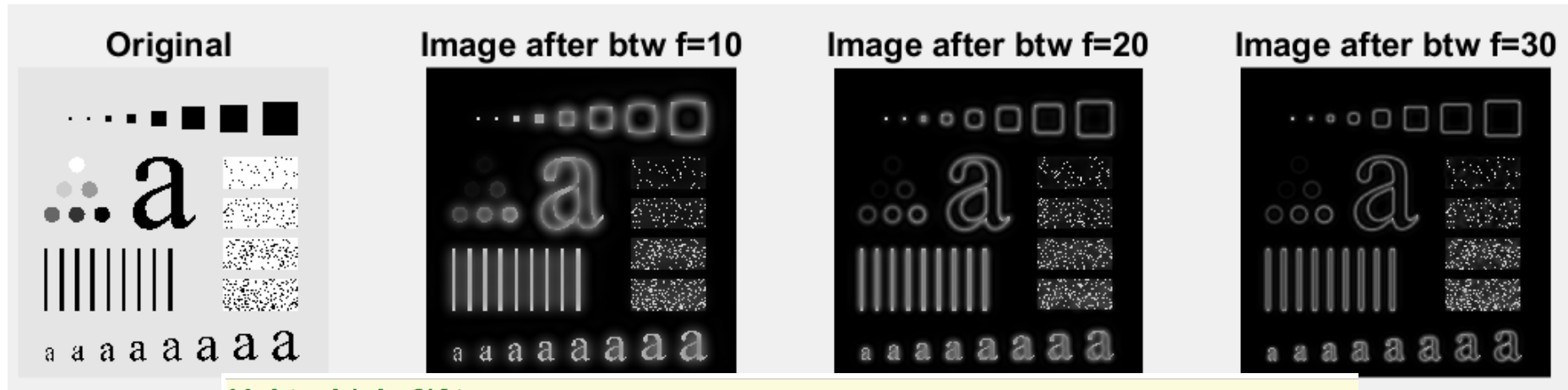
outf=fft2(out);
outf=fftshift(outf);
outf=abs(outf);
outf=log(1+outf);

out1f=fft2(out1);
out1f=fftshift(out1f);
out1f=abs(out1f);
out1f=log(1+out1f);

out2f=fft2(out2);
out2f=fftshift(out2f);
out2f=abs(out2f);
out2f=log(1+out2f);

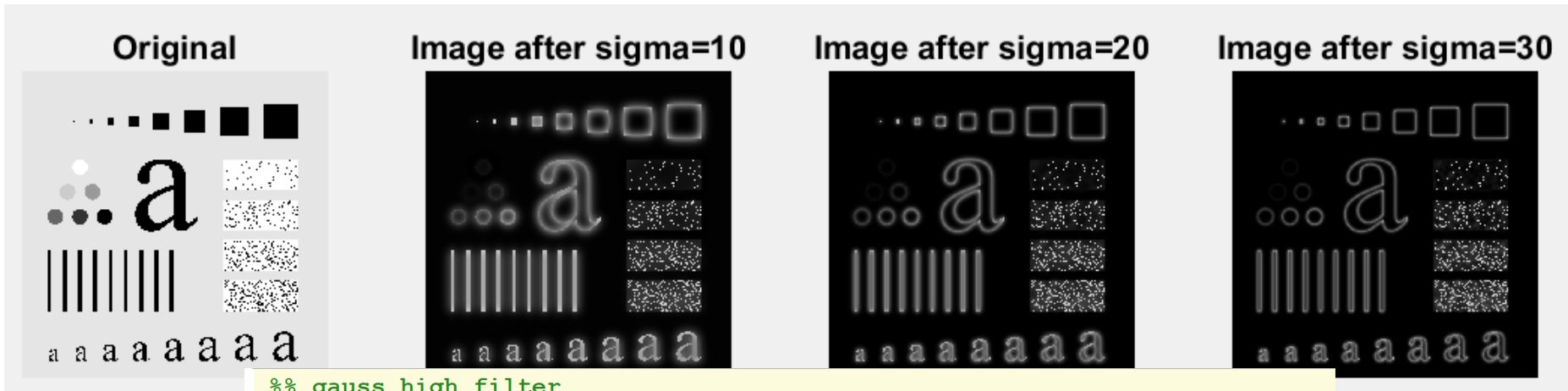
subplot(2,4,5);imshow(If,[]);title('Spectrum');
subplot(2,4,6);imshow(outf,[]);title('Spectrum f=20');
```

Filtering Results by BHPF



```
% bw high filter
I=imread('test1.tif');
sigma1=10;
sigma2=20;
sigma3=30;
ff=imbtwflpf(I,sigma1);ff=1-ff;
out=imfreqfilt(I,ff);
ff1=imbtwflpf(I,sigma2);ff1=1-ff1;
out1=imfreqfilt(I,ff1);
ff2=imbtwflpf(I,sigma3);ff2=1-ff2;
out2=imfreqfilt(I,ff2);
figure;
subplot(1,4,1);imshow(I,[]);title('Original ');
subplot(1,4,2);imshow(mat2gray(abs(out)));title('Image after btw f=10');
subplot(1,4,3);imshow(mat2gray(abs(out1)));title('Image after btw f=20');
subplot(1,4,4);imshow(mat2gray(abs(out2)));title('Image after btw f=30');
```

Filtering Results by GHPF



```
% gauss high filter
clc;
clear;
I=imread('test1.tif');
sigma1=10;
sigma2=20;
sigma3=30;
ff=imgaussflpf(I,sigma1);ff=1-ff;
out=imfreqfilt(I,ff);
ff1=imgaussflpf(I,sigma2);ff1=1-ff1;
out1=imfreqfilt(I,ff1);
ff2=imgaussflpf(I,sigma3);ff2=1-ff2;
out2=imfreqfilt(I,ff2);
figure;
subplot(1,4,1);imshow(I);title('Original');
subplot(1,4,2);imshow(mat2gray(abs(out)));title('Image after sigma=10');
subplot(1,4,3);imshow(mat2gray(abs(out1)));title('Image after sigma=20');
subplot(1,4,4);imshow(mat2gray(abs(out2)));title('Image after sigma=30');
```

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The Laplacian in Spatial Domain

- The Laplacian is defined as follows:

$$\nabla^2 f = \frac{\partial^2 f}{\partial^2 x} + \frac{\partial^2 f}{\partial^2 y}$$

- where the **partial 2nd order derivative** in the x direction is defined as follows:

$$\frac{\partial^2 f}{\partial^2 x} = f(x+1, y) + f(x-1, y) - 2f(x, y)$$

- and in the y direction as follows:

$$\frac{\partial^2 f}{\partial^2 y} = f(x, y+1) + f(x, y-1) - 2f(x, y)$$

The Laplacian in Spatial Domain

- So, the Laplacian can be given as follows:
- $\nabla^2 f = [f(x+1, y) + f(x-1, y) + f(x, y+1) + f(x, y-1) - 4f(x, y)]$
- We can easily build a filter based on this

0	1	0
1	-4	1
0	1	0

Laplacian Filter in Frequency Domain

$$\mathcal{F} \left[\frac{d^n f(x)}{dx^n} \right] = (ju)^n F(u)$$

- The Fourier transform of Laplacian equation is

$$\begin{aligned} \mathcal{F} \left[\frac{\partial^2 f(x,y)}{\partial x^2} + \frac{\partial^2 f(x,y)}{\partial y^2} \right] &= (ju)^2 F(u, v) + (jv)^2 F(u, v) \\ &= -(u^2 + v^2) F(u, v) \end{aligned}$$

$$\mathcal{F}[\nabla^2 f(x, y)] = -(u^2 + v^2) F(u, v)$$

- This means that the Laplacian filter in frequency domain can be express as

$$H(u, v) = -(u^2 + v^2)$$

Basic preperities

Transform	$x(t)$	$\leftrightarrow X(u)$
scaling	$x(at)$	$\leftrightarrow \frac{1}{ a } X\left(\frac{u}{a}\right)$
shift	$x(t - \tau)$	$\leftrightarrow X(u)e^{-j2\pi u\tau}$
modulation	$x(t)e^{j2\pi vt}$	$\leftrightarrow X(u - v)$
scale then shift	$x\left(\frac{t-\tau}{a}\right)$	$\leftrightarrow \frac{1}{ a } X\left(\frac{u}{a}\right)e^{-j2\pi u\tau}$
shift then scale	$x\left(\frac{t}{a} - b\right)$	$\leftrightarrow \frac{1}{ a } X\left(\frac{u}{a}\right)e^{-j2\pi uab}$
derivative	$\left(\frac{d}{dt}\right)^n x(t)$	$\leftrightarrow (j2\pi u)^n X(u)$
integral	$\int_{-\infty}^t x(\tau)d\tau$	$\leftrightarrow \frac{X(u)}{j2\pi u} + \frac{1}{2}X(0)\delta(u)$
conjugate	$x^*(t)$	$\leftrightarrow X^*(-u)$
transpose	$x(-t)$	$\leftrightarrow X(-u)$
inversion	$\int_{-\infty}^{\infty} X(u)e^{j2\pi ut}du$	$\leftrightarrow X(u)$
duality	$X(t)$	$\leftrightarrow x(-u)$
linearity	$ax_1(t) + bx_2(t)$	$\leftrightarrow aX_1(u) + bX_2(u)$
convolution	$x(t) * h(t)$	$\leftrightarrow X(u)H(u)$
correlation	$x(t) \star h(t)$	$\leftrightarrow X(u)H^*(u)$
real signals	if $x(t)$ is real $\Rightarrow R(u) = R(-u),$ $ X(u) = X(-u) ,$	$\leftrightarrow X(u) = X^*(-u),$ $I(u) = -I(-u)$ $\angle\{X(u)\} = -\angle\{X(-u)\}.$
causal signals	$x(t) = x(t)\mu(t)$ $\Rightarrow I(u) = \frac{-j}{\pi u} * R(u),$	$\leftrightarrow X(u) = \frac{-j}{\pi u} * X(u)$ $R(u) = \frac{1}{\pi u} * I(u)$
Fourier series	$\sum_n c_n e^{j2\pi nt/T}$	$\leftrightarrow \sum_n c_n \delta\left(u - \frac{n}{T}\right)$
sampling theorem	$\sum_n x_n \text{sinc}(2Bt - n)$	$\leftrightarrow \sum_n x_n e^{-j\pi u/B} \Pi\left(\frac{u}{2B}\right)$

Laplacian Filter in Frequency Domain

- As the filter origin was shifted to the center of image, we may shift the Laplacian filter in frequency domain by $M/2$ and $N/2$ respectively

$$H(u, v) = -[(u - M / 2)^2 + (v - N / 2)^2]$$

- Dual relationship in the familiar Fourier transform-pair notation:

$$\nabla^2 f(x, y) \Leftrightarrow -[(u - M / 2)^2 + (v - N / 2)^2] F(u, v)$$

Example: Laplacian Filtering

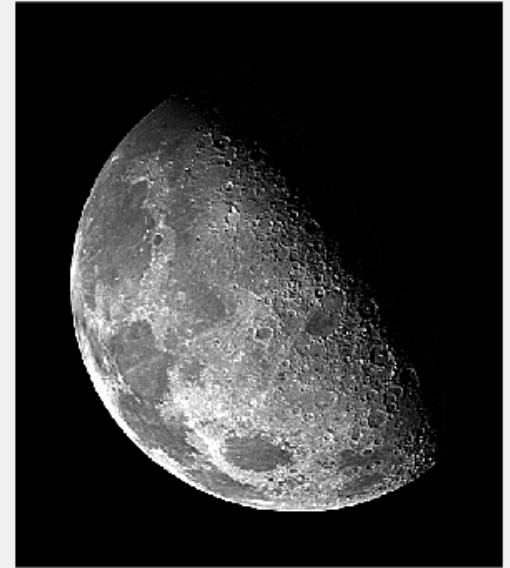
original



Image with Laplacian in Spatial



Image with Laplacian in Frequency



Lecture Outline

- Image Enhancement in Frequency Domain
- Filtering in Frequency Domain
 - Low-Pass Filtering
 - High-Pass Filtering
 - Laplacian Filtering
 - Homomorphic Filtering
 - Selective Filtering – Bandpass/Bandreject, Notch Filters

Homomorphic Filtering

■ Illumination-reflection Model:

- Let $f(x,y)$ be non-zero and finite image that $0 < f(x,y) < \infty$, $f(x,y)$ may be characterized by components:
- Source **illumination** ($i(x,y)$) incident on the scene being viewed
- Amount of **illumination reflection** ($r(x,y)$) by the objects in the scene

■ The two functions combine as a product to form

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

where $0 < i(x,y) < \infty$,

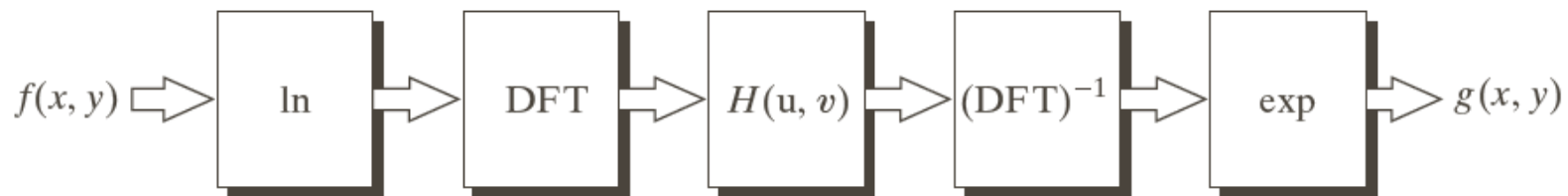
and $0 < r(x,y) < 1$

Homomorphic Filtering

- Illumination and Reflection have different characteristics:
 - Illumination components tend to be slow in spatial variation (**low frequency components**)
 - Reflection of various objects tends to vary abruptly (**high frequency components**)
- Better control can be achieved if the two components are separated by log function and filters are applied separately to each of the respective components.

Homomorphic Filtering

- The whole process can be summarized as follows



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

$$\begin{aligned} Z(u, v) &= \mathfrak{F}\{z(x, y)\} = \mathfrak{F}\{\ln(i(x, y))\} + \mathfrak{F}\{\ln(r(x, y))\} \\ &= F_i(u, v) + F_r(u, v) \end{aligned}$$

$$S(u, v) = H(u, v)Z(u, v) = H(u, v)F_i(u, v) + H(u, v)F_r(u, v)$$

$$s(x, y) = \mathfrak{F}^{-1}\{H(u, v)F_i(u, v)\} + \mathfrak{F}^{-1}\{H(u, v)F_r(u, v)\} = i'(x, y) + r'(x, y)$$

$$g(x, y) = e^{s(x, y)} = e^{i'(x, y)} e^{r'(x, y)} = i_0(x, y) r_0(x, y)$$

Homomorphic Filtering

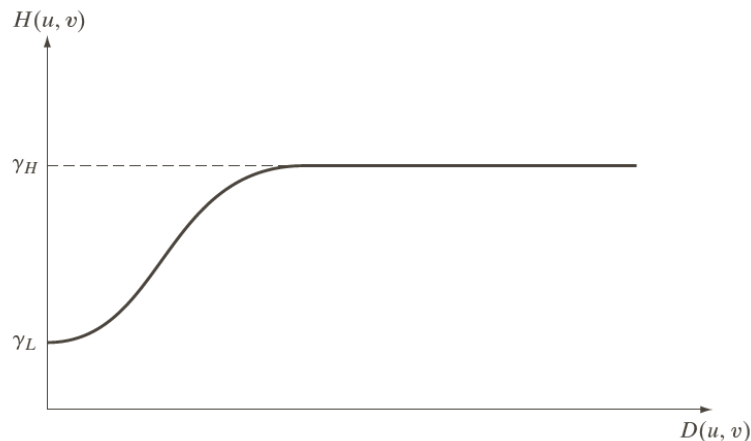
- Filter $H(u,v)$ can be designed such that it tends to **decrease** the contribution made by **low frequencies** (illumination, $\gamma_L < 1$) and **amplify** the contribution made by **high frequencies** (reflectance, $\gamma_H > 1$)
- The result is simultaneous dynamic range compression and contrast enhancement

Homomorphic Filtering

- The curve shape can be approximated using the basic form of any of the ideal highpass filters. Example: using a slightly-modified form of GHPF gives:

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

- where constant c controls the sharpness of the slope of the filter function as it transitions between γ_L and γ_H .



Example: Homomorphic Filtering



```
%% homomorphic
d=10;
order=2;
im=double(imread('PET_image.tif'));
[r, c]=size(im);
homofil(im,d,r,c,order);
```

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Selective Filtering

- The previous filters all operate over the entire frequency rectangle. However, some applications only require **processing of specific bands of frequencies** or small regions of the frequency rectangle
 - Bandreject / Bandpass Filters: process specific bands
 - Notch Filters: process small regions of the frequency rectangle

Selective Filtering:

- Based on previous types of filters (Ideal, Butterworth, Gaussian), **Bandreject** filters can be constructed by adding an addition parameter – width of the band, W

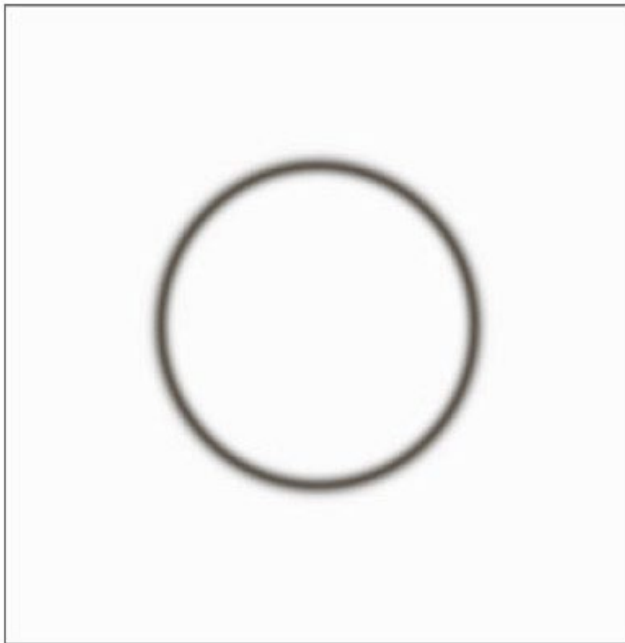
Bandreject filters. W is the width of the band, D is the distance $D(u, v)$ from the center of the filter, D_0 is the cutoff frequency, and n is the order of the Butterworth filter. We show D instead of $D(u, v)$ to simplify the notation in the table.

Ideal	Butterworth	Gaussian
$H(u, v) = \begin{cases} 0 & \text{if } D_0 - \frac{W}{2} \leq D \leq D_0 + \frac{W}{2} \\ 1 & \text{otherwise} \end{cases}$	$H(u, v) = \frac{1}{1 + \left[\frac{DW}{D^2 - D_0^2} \right]^{2n}}$	$H(u, v) = 1 - e^{-\left[\frac{D^2 - D_0^2}{DW} \right]^2}$

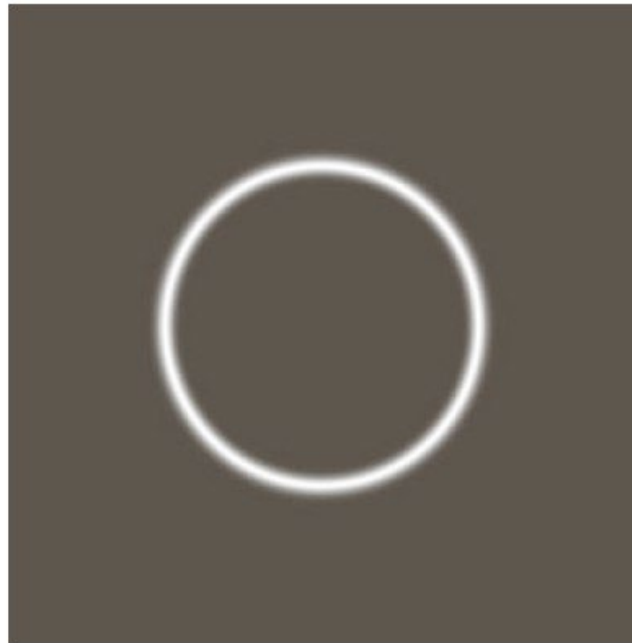
- Equivalent **Bandpass** filters can be obtained from a bandreject filter by inverting its effect

$$H_{BP}(u, v) = 1 - H_{BR}(u, v)$$

Selective Filtering



Bandreject Gaussian filter



Corresponding bandpass filter

Notch Filters

- **A filter that rejects (or passes) specific frequencies**
- Example: periodic noise corresponds to spikes or lines in the Fourier domain
- Can design a filter with zeros at those frequencies, this will remove the noise
- Examples:
 - Image mosaics
 - Scan line noise
 - Halftoning noise (moire patterns)

Steps in Notch Filtering

Look at spectrum $|F(u,v)|$ of noisy image $f(x,y)$, find frequencies corresponding to the noise

Create a mask image $M(u,v)$ with notches (zeros) at those places, 1's elsewhere

Multiply mask with original image transform; this zeros out noise frequencies

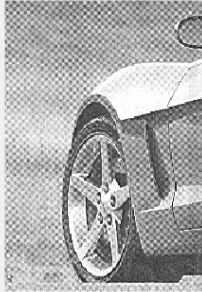
$$G(u,v) = M(u,v) F(u,v)$$

Take inverse Fourier transform to get restored image

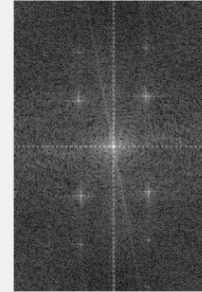
$$g(x,y) = \mathcal{F}^{-1}(G(u,v))$$

Example Notch Filters

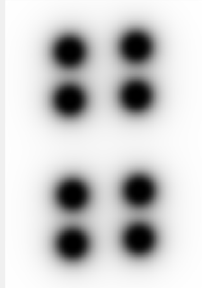
a).Original Image



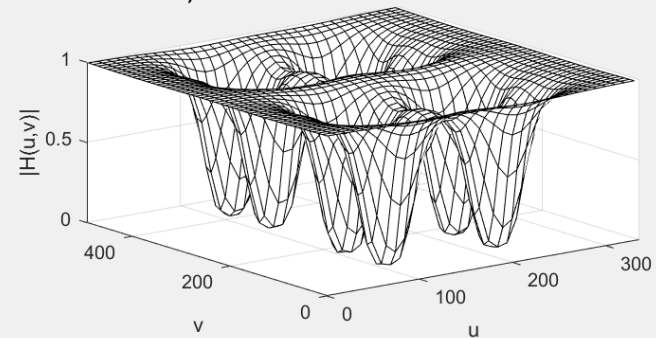
b).Fourier Spectrum



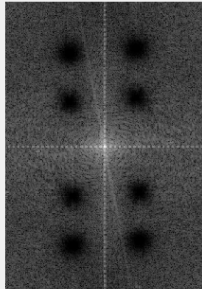
c).Butterworth Notch Filter($D=30$ $n=2$)



d).3D View of Butterworth Notch Filter



d).Fourier Spectrum



e).Result image



Summary

- An “ideal lowpass filter” passes all frequencies with magnitudes below a specific level, and attenuates all frequencies above that level.
- An “ideal highpass filter” does the opposite.
- A “notch” filter rejects (or passes) frequencies at a specific point (the notch).

