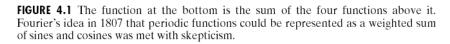


Fourier Transform

•Pourier Series: Any function that periodically repeats itself can be expressed which will be as the sum of sines/cosines of different frequencies, each multiplied with a different coefficient.

•Fourier Transform: Functions that are not periodic, whose area under the curve is finite, can be expressed as the integral of sines and/cosines multiplied by a weighting function.





1D Continuous Fourier Transform

• The Fourier Transform is an important tool in Image Processing, and is directly related to filter theory, since a filter, which is a <u>convolution</u> in the spatial domain, is a simple <u>multiplication</u> in the frequency domain.

•1-D Continuous Fourier Transform

The Fourier transform, F(u), of a single variable continuous function, f(x), is defined by:

$$F(u) = \int_{-\infty}^{\infty} f(x)e^{-j2\pi ux}dx$$

Given Fourier transform of a function F(u), the inverse Fourier transform can be used to obtain, f(x), by:

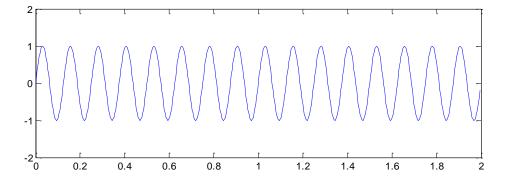
$$f(x) = \int_{-\infty}^{\infty} F(u)e^{j2\pi ux}du$$

Note: F(u), is the frequency spectrum, where, u represents the frequency, and f(x) is the signal where x represents time/space. $j = \sqrt{-1}$ Prepared By: Dr. Hasan Demirel, PhD



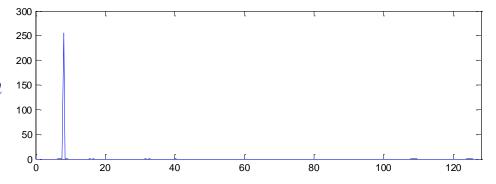
1D Continuous Fourier Transform

Time/Space domain



Sine wave

Frequency domain

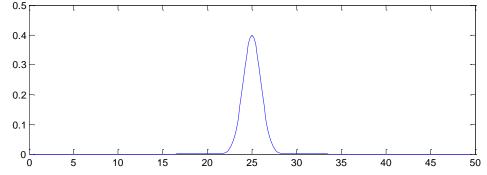


Delta function



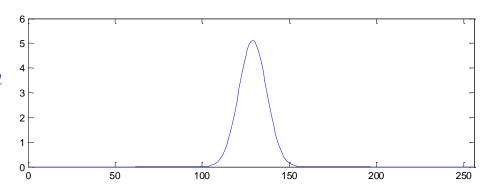
1D Continuous Fourier Transform





Gaussian

Frequency domain

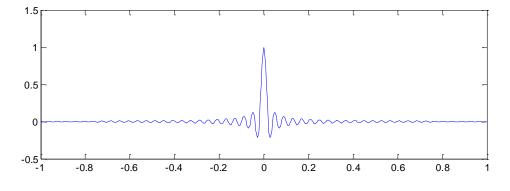


Gaussian



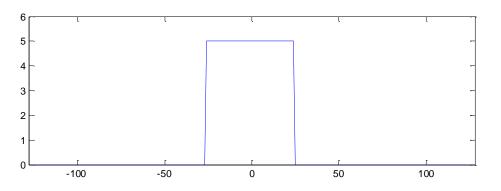
10 Continuous Fourier Transform





Sinc function

Frequency domain



Square wave

Fourier Transform pairs (spatial versus Frequency)

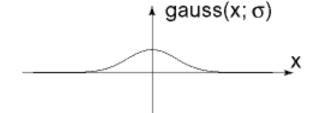
Spatial domain

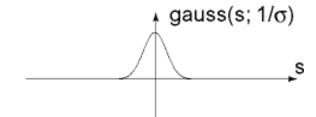
$$f(x)$$
 $box(x)$

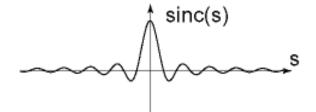
Frequency domain

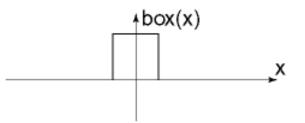
$$F(s) = \int_{-\infty}^{\infty} f(x)e^{-i2\pi sx}dx$$

$$sinc(s)$$













1D Discrete Fourier Transform

• I-D Discrete Fourier Transform

The Fourier transform, F(u), of a discrete function of one variable, f(x), x=0, 1, 2, ..., M-1, is given by:

$$F(u) = \frac{1}{M} \sum_{x=0}^{M-1} f(x) e^{-j2\pi ux/M}$$

The inverse Discrete Fourier Transform can be used to calculate f(x), by:

$$f(x) = \sum_{u=0}^{M-1} F(u)e^{j2\pi ux/M}$$

Note: F(u), which is the Fourier transform of f(x) contains discrete complex quantities and it has the same number of components as f(x).

$$e^{j\theta} = \cos\theta + j\sin\theta$$





1D Discrete Fourier Transform

•1-D Discrete Fourier Transform

Then;

$$F(u) = \frac{1}{M} \sum_{x=0}^{M-1} f(x) [\cos 2\pi u x / M - j \sin 2\pi u x / M]$$

•The Fourier Transform generates complex quantities. The magnitude or the spectrum of the Fourier transform is given by:

$$|F(u)| = \left[R^2(u) + I^2(u)\right]^{1/2}$$
 $R(u)$ is the Real Part and $I(u)$ is the Imaginary Part

•The Phase Spectrum of the transform refers to the angles between the real and imaginary components and it is denoted by:

$$\phi(u) = \tan^{-1} \left[\frac{I(u)}{R(u)} \right]$$





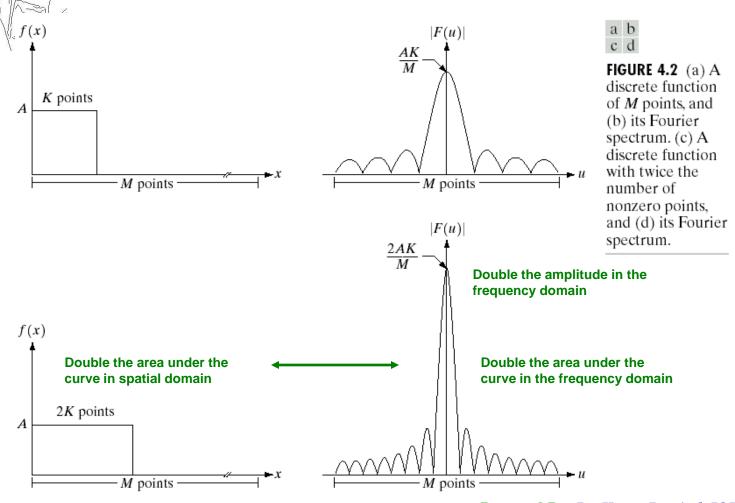
1D Discrete Fourier Transform

- I Discrete Fourier Transform
- •The Power Spectrum/spectral density is defined as the square of the Fourier spectrum and denoted by

$$P(u) = |F(u)|^2 = R^2(u) + I^2(u)$$

- •The power spectrum can be used, for example to separate a portion of a specified frequency (i.e. low frequency) power from the power spectrum and monitor the effect. Thypically used to define the cut off frequencies used in lowpass and higpass filtering.
- We primarily use the Fourier Spectrum for image enhancement applications.

1D Discrete Fourier Transform







2D Discrete Fourier Transform

The Fourier transform of a 2D discrete function (image) f(x,y) of size MxN is given by:

$$F(u,v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x,y) e^{-j2\pi(ux/M + vy/N)}$$

u=0, 1, 2, ..., M-1, and v=0, 1, 2, ..., N-1 and the inverse 2D Discrete Fourier Transform can be calculated by:

$$f(x,y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} f(u,v)e^{j2\pi(ux/M + vy/N)}$$

x=0, 1, 2, ..., M-1, and y=0, 1, 2, ..., N-1.



2D Discrete Fourier Transform

The 2D Fourier Spectrum, Phase Spectrum and Power Spectrum can be respectively denoted by:

$$|F(u,v)| = [R^2(u,v) + I^2(u,v)]^{1/2}$$
 Magnitude/Fourier Spectrum

$$\phi(u, v) = \tan^{-1} \left[\frac{I(u, v)}{R(u, v)} \right]$$

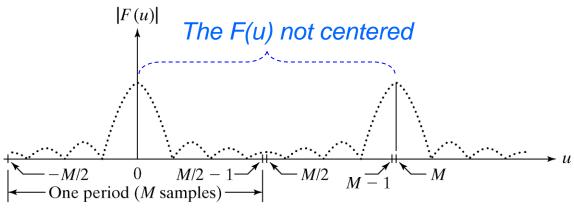
Phase Spectrum

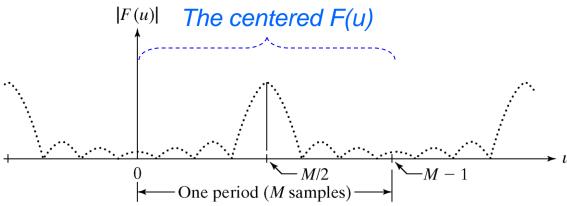
$$P(u,v) = |F(u,v)|^2 = R^2(u,v) + I^2(u,v)$$
 Power Spectrum



2D Discrete Fourier Transform

The Periodicity property: F(u) in 1D DFT has a period of M
The Symmetry property: The magnitude of the transform is centered on the origin.





a b

FIGURE 4.1

- (a) Fourier spectrum showing back-to-back half periods in the interval [0 M 1]
- [0, M-1].
- (b) Centered spectrum in the same interval, obtained by multiplying f(x) by $(-1)^x$ prior to computing the Fourier transform.

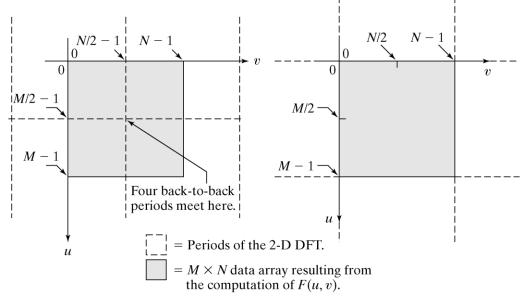




2D Discrete Fourier Transform

The $\frac{\textbf{Periodicity}}{\textbf{property}}$ property: F(u,v) in 2D DFT has a period of N in horizontal and M in vertical directions

The Symmetry property: The magnitude of the transform is centered on the origin.



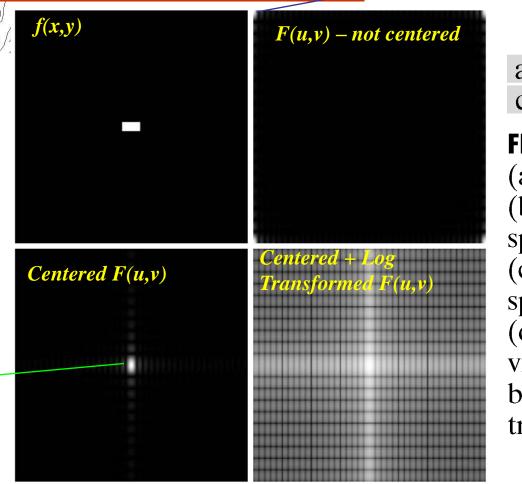
a b

FIGURE 4.2 (a) $M \times N$ Fourier spectrum (shaded), showing four back-to-back quarter periods contained in the spectrum data. (b) Spectrum obtained by multiplying f(x, y) by $(-1)^{x+y}$ prior to computing the Fourier transform. Only one period is shown shaded because this is the data that would be obtained by an implementation of the equation for F(u, v).

2D Discrete Fourier Transform

Origin = $0.0 \leftarrow$

Origin = 0,0



a b c d

FIGURE 4.3

- (a) A simple image.
- (b) Fourier spectrum.
- (c) Centered spectrum.
- (d) Spectrum visually enhanced by a log transformation.

EE-583: Digital Image Processing

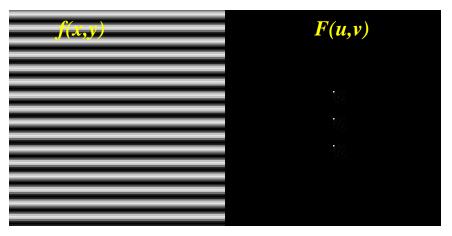
Image Enhancement in the Frequency Domain

2D/Discrete Fourier Transform

Consider the following 2 left images which are pure vertical cosine of 4 cycles and a pure vertical cosine of 32 cycles.



Notice that the Fourier Spectrum for each image at the right contains just a single component, represented by 2 bright spots symmetrically placed about the center.



The dot at the center that represents the (0,0) frequency term or average value of the image. Images usually have a large average value/DC component. Due to low frequencies Fourier Spectrum images usually have a bright blob of components near the center.

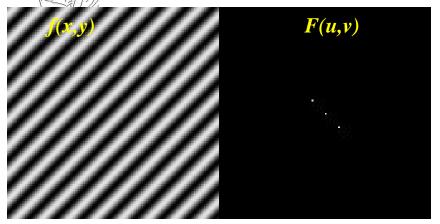
Prepared By: Dr. Hasan Demirel, PhD



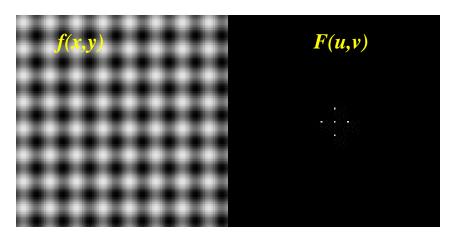


2D Discrete Fourier Transform

Consider the following 2 left images with pure cosines in pure forward diagonal and mixed vertical horizontal.



One of the properties of the 2D FT is that if you rotate the image, the spectrum will rotate in the same direction



The sum of 2 sine functions, each in opposite (vertical and horizontal) direction.

EE-583: Digital Image Processing

Image Enhancement in the Frequency Domain

2D Discrete Fourier Transform

The center value (at the origin) of the Frequency Spectrum corresponds to the <u>ZERO</u> frequency component which also referred to as the <u>DC</u> component in an image:

Substituting 0,0 to the origin, the Fourier transform function yields to the average/DC

component value as follows:

 $F(0,0) = \frac{1}{MN} \sum_{x=0}^{M} \sum_{y=0}^{M} f(x,y)$

f(x,y) F(u,v)

f(x,y) (Average image)

The center/zero frequency component (DC component) is removed

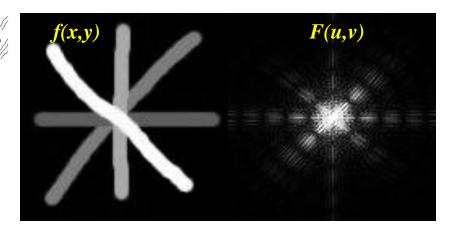
F(u,v)
(DC Component)

Prepared By: Dr. Hasan Demirel, PhD

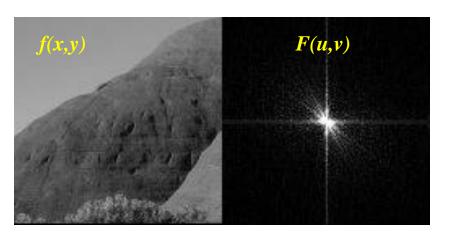
f(x,y)



2D Discrete Fourier Transform



The lines in an image often generate perpendicular lines in the spectrum



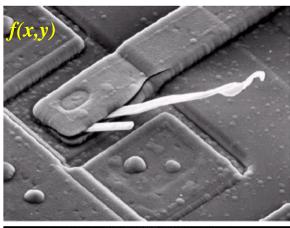
The sloped lines in the spectrum are due to the sharp transition from the sky to the mountain

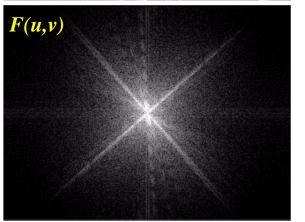




Some Basic Properties of the Frequency Domain:

Frequency is directly related to the rate of change. Therefore, slowest varying component (u=v=0) corresponds to the average intensity level of the image. Corresponds to the origin of the Fourier Spectrum.





a b

FIGURE 4.4

(a) SEM image of a damaged integrated circuit. (b) Fourier spectrum of (a). (Original image courtesy of Dr. J. M. Hudak, Brockhouse Institute for Materials Research, McMaster University,

Hamilton, Ontario, Canada.) •Higher frequencies corresponds to the faster varying intensity level changes in the image. The edges of objects or the other components characterized by the abrupt changes in the intensity level corresponds to higher frequencies.



Basic Steps for Filtering in the Frequency Domain:

Frequency domain filtering operation

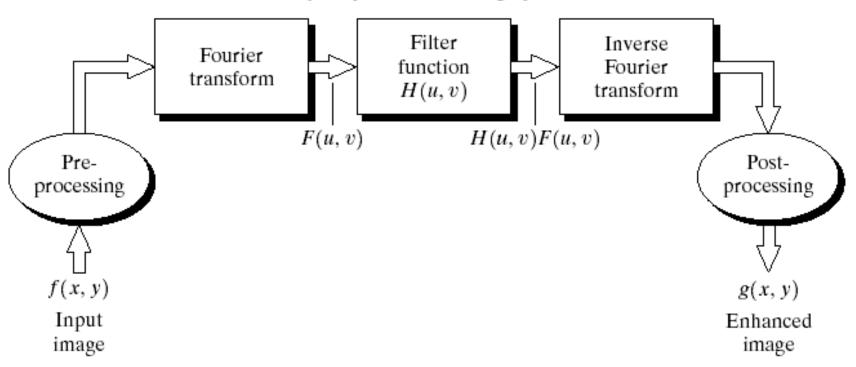


FIGURE 4.5 Basic steps for filtering in the frequency domain.





Image Enhancement in the Frequency Domain Filtering in the Frequency Domain

•Basic Steps for Filtering in the Frequency Domain:

- 1. Multiply the input image by $(-1)^{x+y}$ to center the transform.
- 2. Compute F(u,v), the DFT of the image from (1).
- 3. Multiply F(u,v) by a filter function H(u,v).
- 4. Compute the inverse DFT of the result in (3).
- 5. Obtain the real part of the result in (4).
- 6. Multiply the result in (5) by $(-1)^{x+y}$.

Given the filter H(u,v) (filter transfer function) in the frequency domain, the Fourier transform of the output image (filtered image) is given by:

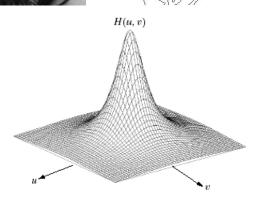
$$G(u,v) = H(u,v)F(u,v)$$
 Step (3)

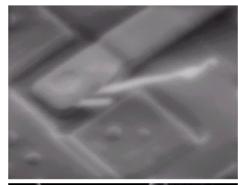
The filtered image g(x,y) is simply the inverse Fourier transform of G(u,v).

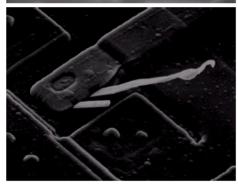
$$g(x, y) = \mathfrak{I}^{-1} [G(u, v)]$$
 Step (4)



•Basics of Low Pass Filters in the Frequency Domain:







- •lowpass filter: A filter that attenuates high frequencies while passing the low frequencies.
- •Low frequencies represent the graylevel appearance of an image over smooth areas.
- •highpass filter: A filter that attenuates low frequencies while passing the high frequencies.
- •High frequencies represents the details such as edges and noise.



H(u, v)

Origin





Filtering in the Frequency Domain

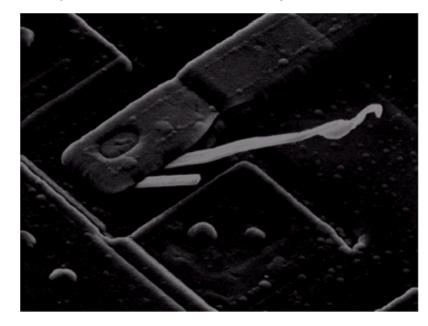
•Consider the following filter transfer function:

$$H(u,v) = \begin{cases} 0 & \text{if } (u,v) = (M/2, N/2) \\ 1 & \text{otherwise} \end{cases}$$

•This filter will set F(0,0) to zero and leave all the other frequency components. Such a filter is called the notch filter, since it is constant function with a hole (notch) at the origin.

FIGURE 4.6

Result of filtering the image in Fig. 4.4(a) with a notch filter that set to 0 the F(0,0) term in the Fourier transform.



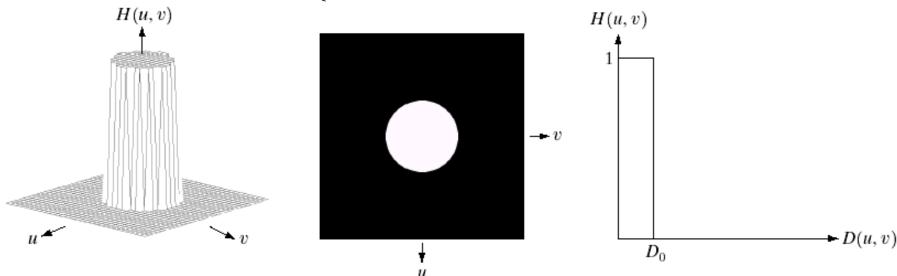


Filtering in the Frequency Domain

Smoothing Frequency Domain Filters: <u>Ideal</u> Lowpass Filters

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency



abc

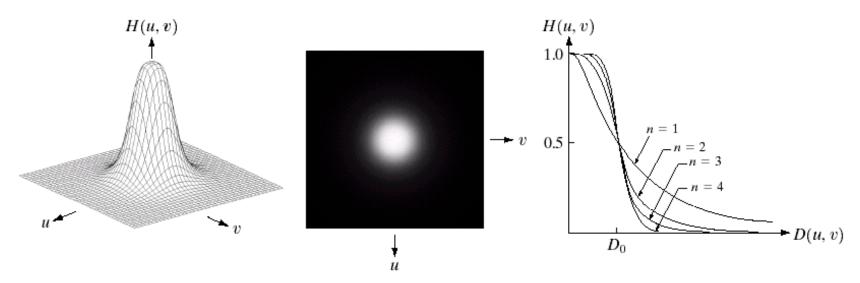
FIGURE 4.10 (a) Perspective plot of an ideal lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross section.

Filtering in the Frequency Domain

Smoothing Frequency Domain Filters: Butterworth Lowpass Filters

$$H(u,v) = \frac{1}{1 + [D(u,v)/D_0]^{2n}} \quad \begin{array}{l} D(u,v) \text{ is the distance from the origin} \\ D_0 \text{ is the order of the filter.} \end{array}$$

n is the order of the filter



a b c

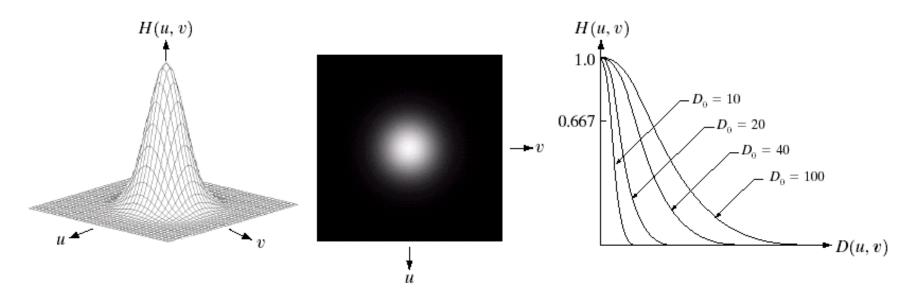
FIGURE 4.14 (a) Perspective plot of a Butterworth lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections of orders 1 through 4.

Filtering in the Frequency Domain

Smoothing Frequency Domain Filters: Gaussian Lowpass Filters

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency.



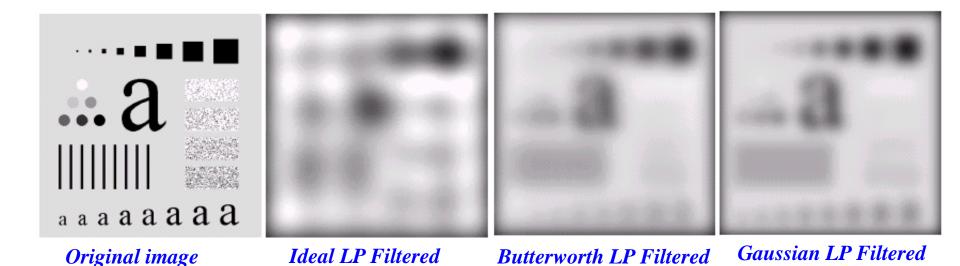
a b c

FIGURE 4.17 (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 .



Riltering in the Frequency Domain

comparison of Ideal, Butterworth and Gaussian Lowpass Filters having the same radii (cutoff frequency) value.







- The high-frequency components are: edges and sharp transitions such as noise.
- •Smoothing/blurring can be achieved by attenuating a specified range of high-frequency components in the frequency domain.
- •Given the Fourier transformed image F(u), the filtered image G(u) can be obtained by:

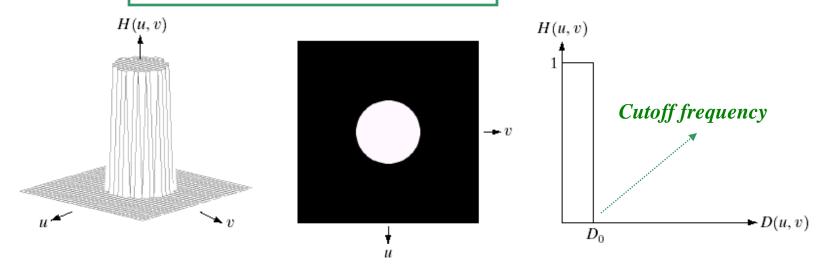
$$G(u,v) = H(u,v)F(u,v)$$

- •Where H(u) is the filter transfer function.
- Smoothing can be achieved by lowpass filters. We will consider only 3 types of lowpass filters:
 - •Ideal Lowpass filters,
 - •Butterworth Lowpass filters,
 - •Gaussian Lowpass filters

• Ideal Lowpass Filter (ILPF): Simply cuts off all the high frequencies higher than the specified cutoff frequency. The filter transfer function:

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency



a b c

FIGURE 4.10 (a) Perspective plot of an ideal lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross section.



- Cutoff Frequency of an Ideal Lowpass Filter: One way of specifying the cutoff frequency is to compute circles enclosing specified amounts of total image power.
- •Calculating P_T which is the total power of the transformed image:

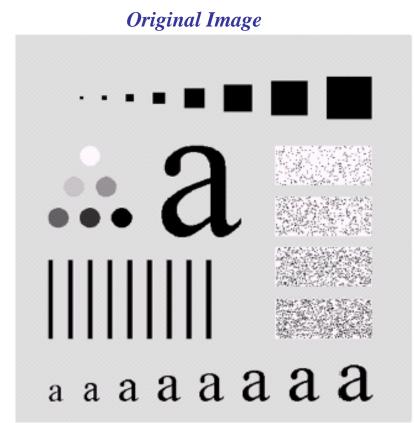
$$P_T = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} P(u,v) \qquad u = 0, 1, ..., M-1 \quad , v = 0, 1, ..., N-1$$

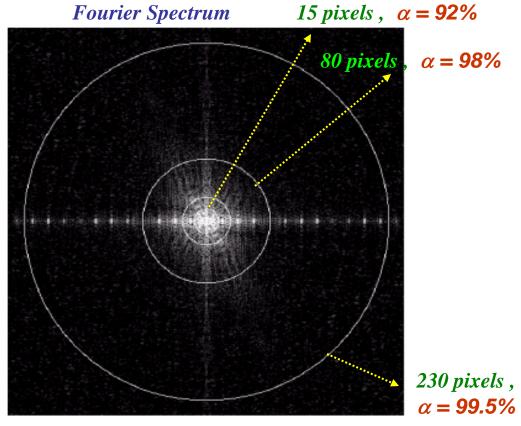
•The cutoff frequency D_o can be determined by specifying the α percent of the total power enclosed by a circle centered at the origin. The radius of the circle is the cutoff frequency D_o .

$$\alpha = 100 \left[\sum_{u} \sum_{v} P(u, v) / P_{T} \right] \qquad u \leq radius(D_{o}) \quad , v \leq radius(D_{o})$$



Outoff Frequency of an Ideal Lowpass Filter:



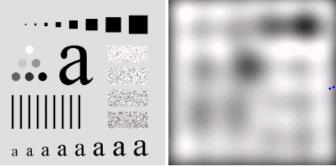




Smoothing Frequency-Domain Filters

Cutoff Frequency of an Ideal Lowpass Filter:

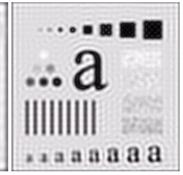
Original Image



ILPF with cutoff=5, removed power=8%

Blurring effect is the result of LPIF

cutoff=15, r.d. power=5.4%

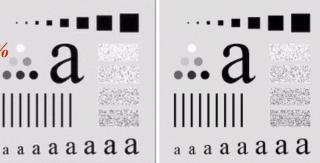


ILPF with cutoff=30, removed power=3.6%

Ringing effect is the problem of LPIF

ILPF with cutoff=230, removed power=0.5%

cutoff=80, r.d. power=2%





Smoothing Frequency-Domain Filters

- Blurring and Ringing properties of ILPF:
- •The blurring and ringing properties of the ILPF can be explained by the help of the convolution theorem:
- •Given the Fourier transformed input image F(u) and output image G(u) and the filter transfer function H(u),

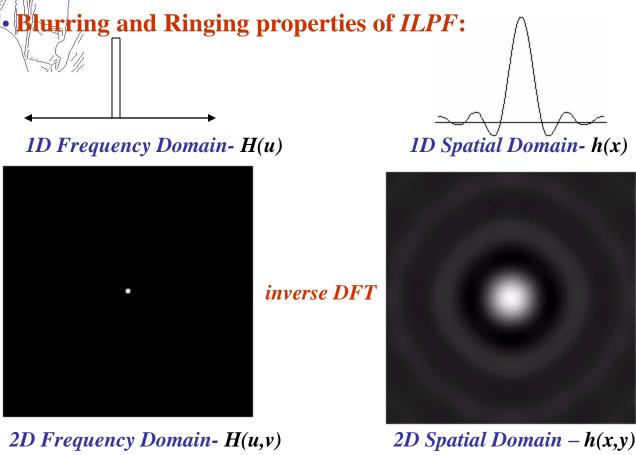
$$G(u,v) = H(u,v)F(u,v)$$

•The corresponding process in the spatial domain with regard to the convolution theorem is given by:

$$g(x, y) = h(x, y) * f(x, y)$$

- •h(x,y) in the spatial domain, is the inverse Fourier transform of the filter transfer function H(u,v).
- The spatial filter h(x,y) has two major characteristics:
 - dominant component at the origin
 - •Concentric circular components about center component.

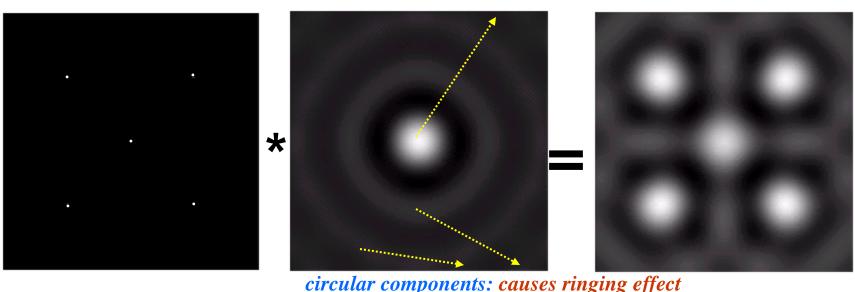
Smoothing Frequency-Domain Filters



- •The spatial domain filters center component is responsible for blurring.
- •The circular components are responsible for the ringing artifacts.

• Blurring and Ringing properties of ILPF: Lets consider the following convolution in the spatial domain:

Central components: causes blurring



Input image - f(x,y)

Spatial Filter – h(x,y)

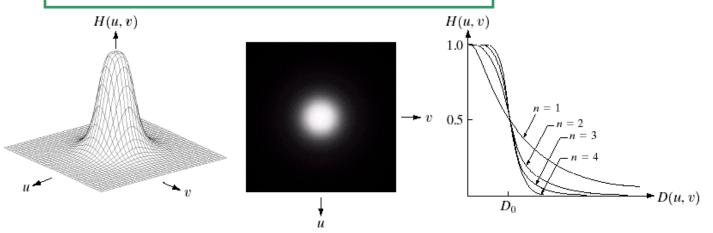
filtered image -g(x,y)



• Butterworth Lowpass Filter (BLPF): The transfer function of BLPF of order n and with a specified cutoff frequency is denoted by the following filter transfer function:

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

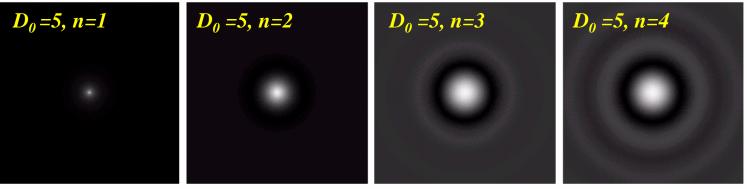
D(u,v) is the distance from the origin D_0 is the cutoff frequency. n is the order of the filter



a b c

FIGURE 4.14 (a) Perspective plot of a Butterworth lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections of orders 1 through 4.

- Butterworth Lowpass Filter (BLPF):
- The BLPF with order of 1 does not have any ringing artifact.
- •BLPF with orders 2 or more shows increasing ringing effects as the order increases.







1D - Profile through the centre of the Spatial domain filters.

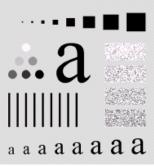


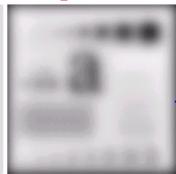
Image Enhancement in the Frequency Domain

Smoothing Frequency-Domain Filters

Butterworth Lowpass Filter (BLPF):

Original Image

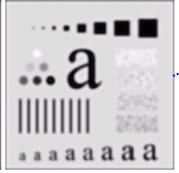




BLPF with cutoff=5, order=2

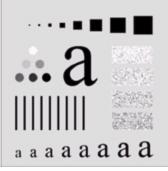
cutoff=15,
order=2

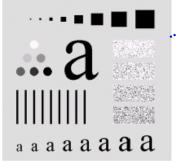




BLPF with cutoff=30, order=2

cutoff=80,
order=2



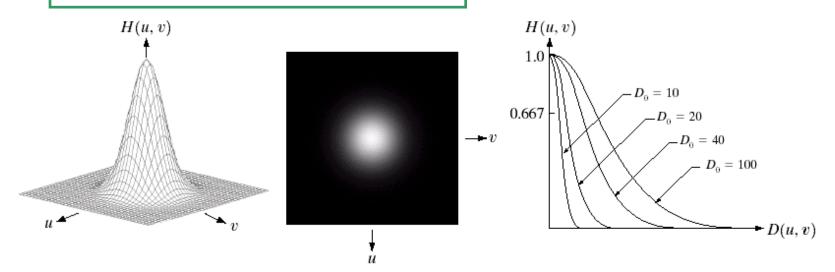


BLPF with cutoff=230, order=2

• Gaussian Lowpass Filter (GLPF): The transfer function of GLPF is given as follows:

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency.



a b c

FIGURE 4.17 (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 .



Smoothing Frequency-Domain Filters





cutoff=80









• (GLPF): The inverse Fourier transform of the Gaussian Lowpass filter is also Gaussian in the Spatial domain.

•Therefore there is no ringing effect of the GLPF. Ringing artifacts are cutoff=30 not acceptable in fields like medical imaging. Hence use Gaussian instead of the ILPF/BLPF.

cutoff=230



• Caussian Lowpass Filter (GLPF): Refer to the improvement in the following example.

a b

FIGURE 4.19

(a) Sample text of poor resolution (note broken characters in magnified view). (b) Result of filtering with a GLPF (broken character segments were joined).

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.





• Caussian Lowpass Filter (GLPF): The following example shows a lady younger, How? By using GLPF!



a b c

FIGURE 4.20 (a) Original image (1028 \times 732 pixels). (b) Result of filtering with a GLPF with $D_0 = 100$. (c) Result of filtering with a GLPF with $D_0 = 80$. Note reduction in skin fine lines in the magnified sections of (b) and (c).





• The high-frequency components are: edges and sharp transitions such as noise.

•Sharpening can be achieved by highpass filtering process, which attenuates low frequency components without disturbing the high-frequency information in the frequency domain.

•The filter transfer function, $H_{hp}(u,v)$, of a highpass filter is given by:

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

- •Where $H_{lp}(u,v)$, is the transfer function of the corresponding lowpass filter.
- We will consider only 3 types of sharpening highpass filters:
 - •Ideal Highpass filters,
 - •Butterworth Highpass filters,
 - •Gaussian Highpass filters

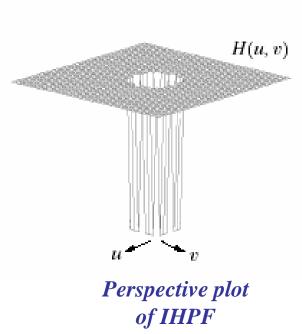


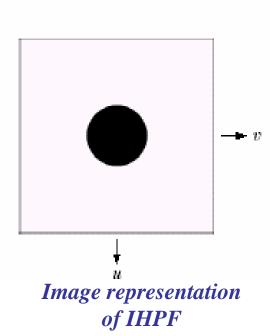
Sharpening Frequency-Domain Filters

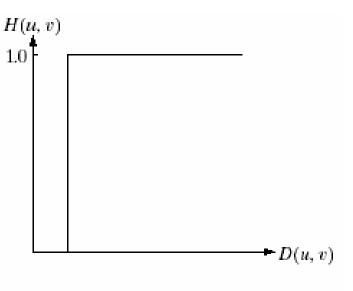
• Ideal Highpass Filter (IHPF): Simply cuts off all the low frequencies lower than the specified cutoff frequency. The filter transfer function:

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency





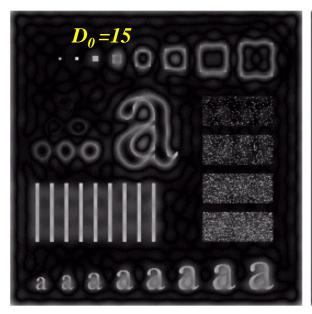


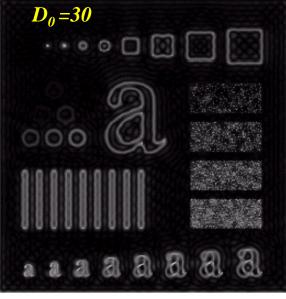
Cross section of IHPF

Prepared By: Dr. Hasan Demirel, PhD



•Ideal Highpass Filter (IHPF): The ringing artifacts occur at low cutoff frequencies





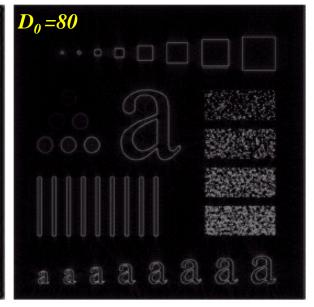




Image Enhancement in the Frequency Domain

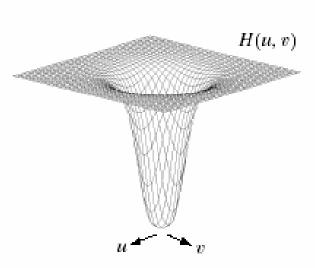
Sharpening Frequency-Domain Filters

• Butterworth Highpass Filter (BHPF): The transfer function of BHPF of order n and with a specified cutoff frequency is given by:

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency.

n is the order of the filter



Perspective plot of BHPF

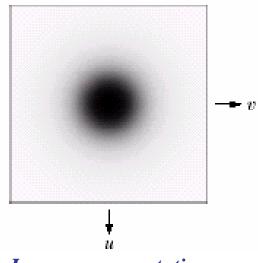
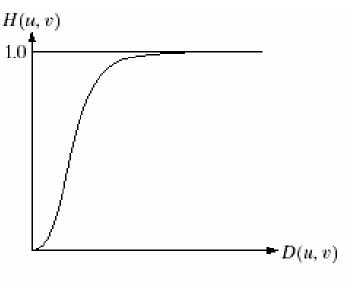


Image representation of BHPF

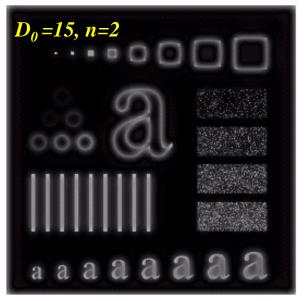


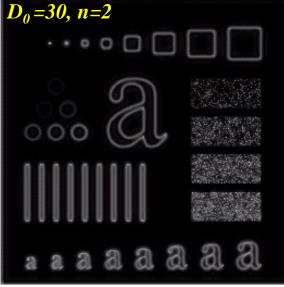
Cross section of BHPF

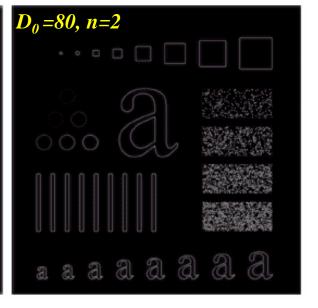
Prepared By: Dr. Hasan Demirel, PhD



*Butterworth Highpass Filter (BHPF): Smoother results are obtained in BIPF when compared IHPF. There is almost no ringing artifacts.





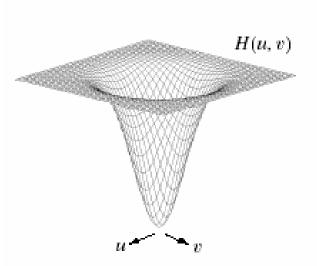




• Caussian Highpass Filter (GHPF): The transfer function of GHPF is given

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

D(u,v) is the distance from the origin D_0 is the cutoff frequency.



Perspective plot of BHPF

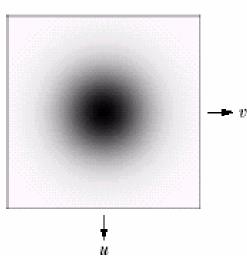
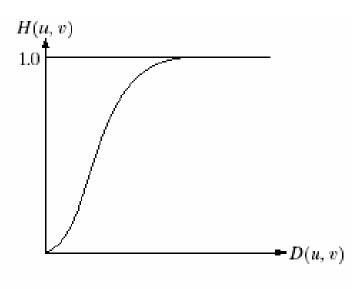


Image representation of BHPF

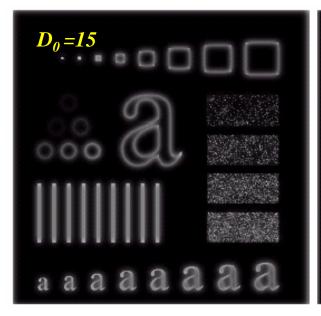


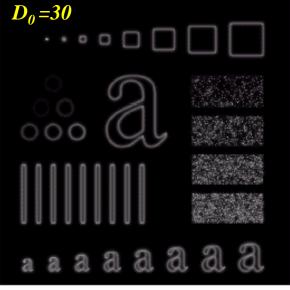
Cross section of BHPF

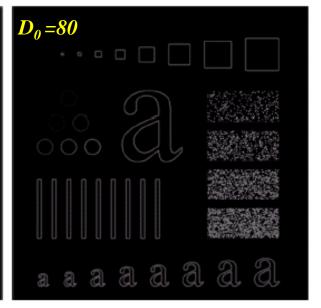
Prepared By: Dr. Hasan Demirel, PhD



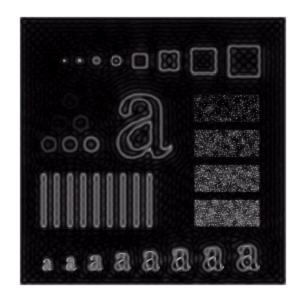
•Gaussian Highpass Filter (GHPF): Smoother results are obtained in GHPF when compared BHPF. There is absolutely no ringing artifacts.



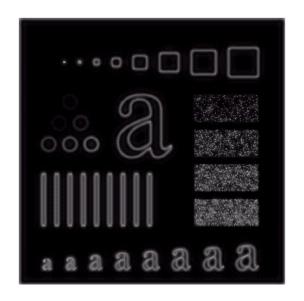




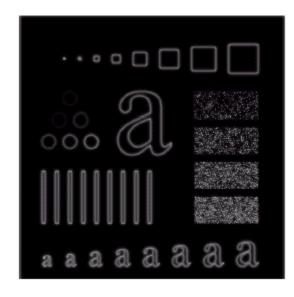
•Comparison of Ideal, Butterworth and Gaussian High pass Filters $(D_0 = 30)$



Ideal High pass Filter



Butterworth High pass Filter



Gaussian High pass Filter