

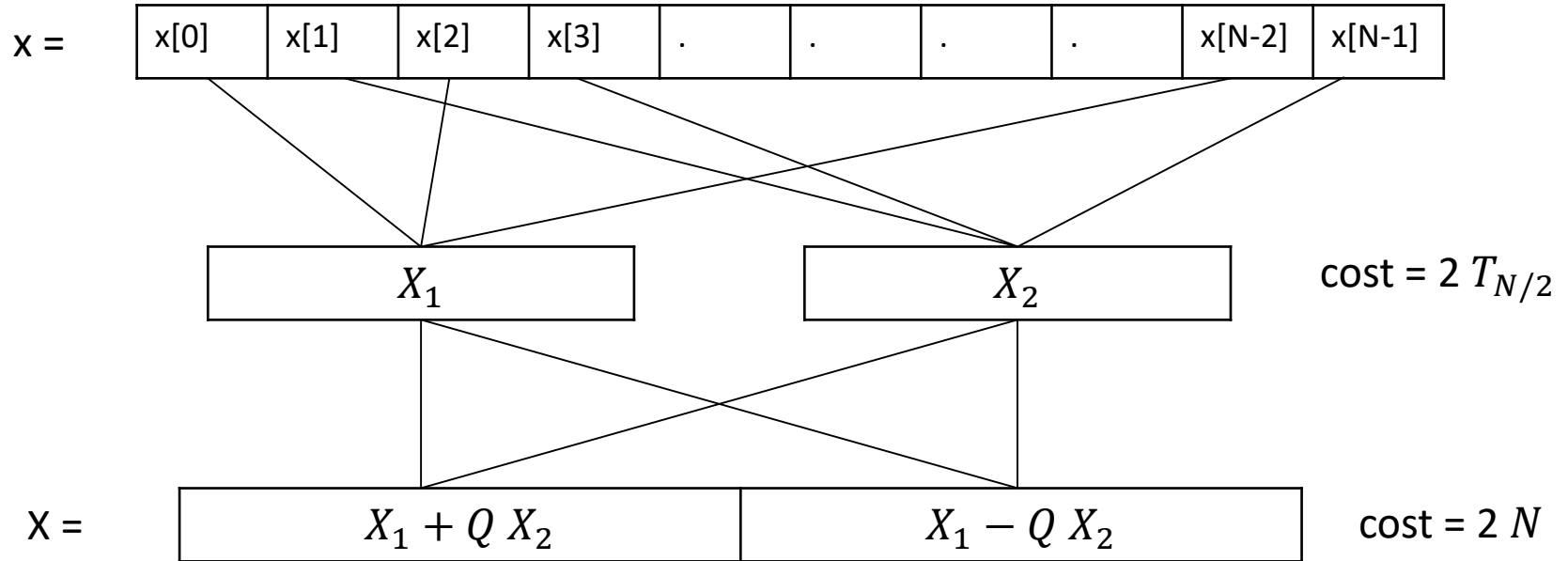
Digital Image Processing (CSE/ECE 478)

Lecture9: Fast Fourier Transforms and Filtering in Fourier Domain

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Center for Visual Information Technology (CVIT), IIIT Hyderabad

Fast Fourier Transform



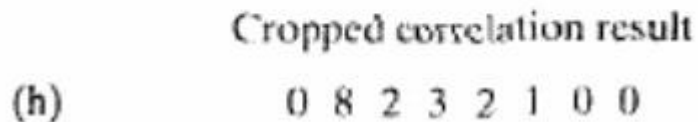
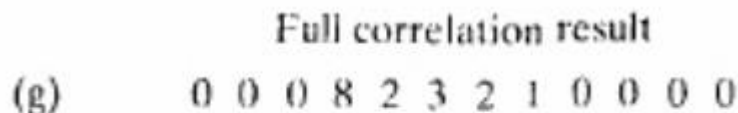
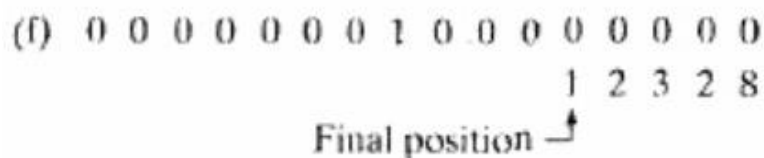
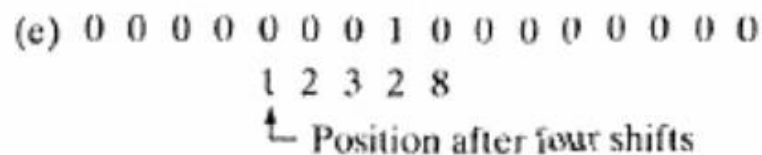
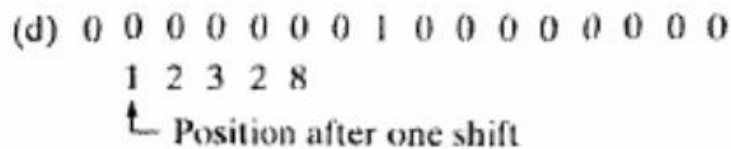
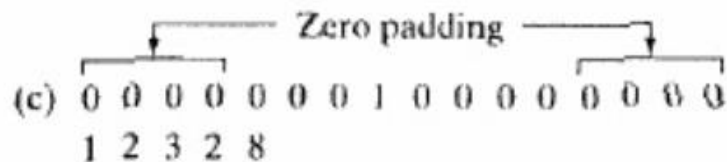
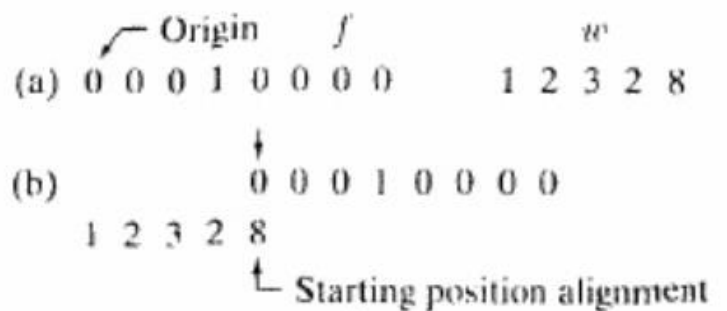
DFT vs FFT computation times

n	$N = 2^n$	N^2	$N \log N$
10	1 024	1 048 576	10 240
12	4 096	16 777 216	49 152
14	16 384	268 435 456	229 376
16	65 536	4 294 967 296	1 048 576

Today's class

- Convolution Theorem
- Frequency domain filtering
 - Low pass
 - High Pass
 - Laplacian

Correlation



Convolution

Origin	f	w rotated 180°
0 0 0 1 0 0 0 0		8 2 3 2 1 (i)

$$\begin{array}{ccccccccccccccccc} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ & & & & & & & 8 & 2 & 3 & 2 & 1 \end{array}$$
$$\begin{array}{cccccccc} & & & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 8 & 2 & 3 & 2 & 1 & & & & & & \end{array} \quad (j)$$

0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 (n)
8 2 3 2 1

0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 (k)
8 2 3 2 1

Full convolution result

0 0 0 1 2 3 2 8 0 0 0 0 (o)

0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 (I)
8 2 3 2 1

Cropped convolution result

0	1	2	3	2	8	0	0
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(p)

Convolution vs Correlation (2D)

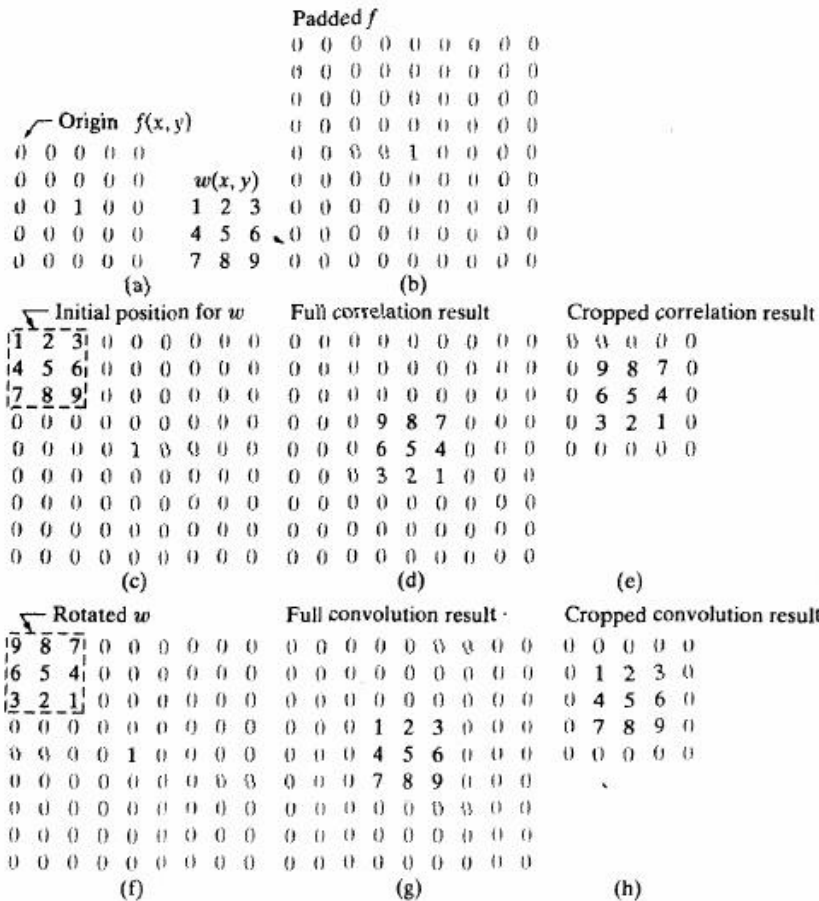


FIGURE 3.30
Correlation (middle row) and convolution (last row) of a 2-D filter with a 2-D discrete, unit impulse. The 0s are shown in gray to simplify visual analysis.

Convolution (2D)

$$w(x, y) \star f(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x - s, y - t)$$

- Evaluated for all values of displacement variables x and y
- Filter size $m \times n$ (notational convenience $\rightarrow m, n$ are assumed odd)
- $a = (m-1)/2$ and $b = (n-1)/2$

Convolution Theorem

$$f(x, y) \star h(x, y) \Leftrightarrow F(u, v)H(u, v)$$

In other words:

$$\mathfrak{F}(f(x, y) \star h(x, y)) = F(u, v)H(u, v)$$

$$f(x, y) \star h(x, y) = \mathfrak{F}^{-1}(F(u, v)H(u, v))$$

Correspondence to spatial filtering



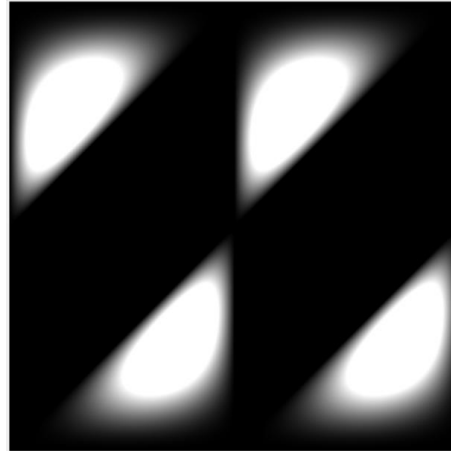
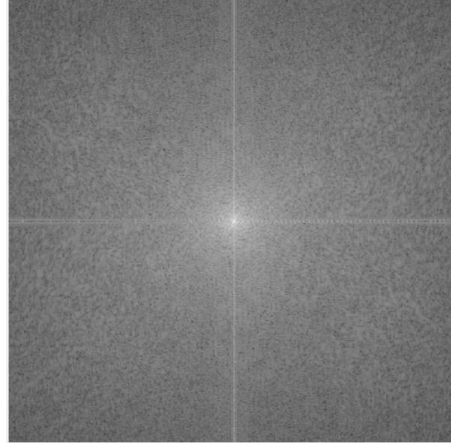
-1	0	1
-2	0	2
-1	0	1



Correspondence to spatial filtering



-1	0	1
-2	0	2
-1	0	1

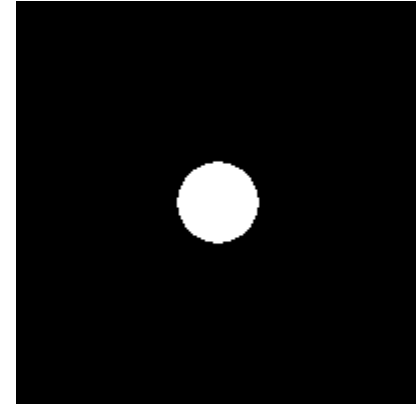
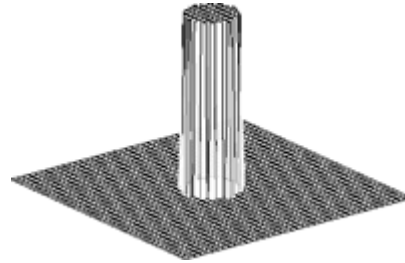


Correspondence to spatial filtering

```
%Sobel filter in frequency domain  
f = rgb2gray(imread('boy.jpg'));  
h = [-1 0 1; -2 0 2; -1 0 1];  
F = fft2(double(f), 402, 402);  
H = fft2(double(h), 402, 402);  
F_fH = fftshift(H).*fftshift(F);  
ffi = ifft2(ifftshift(F_fH));
```

Ideal Low Pass Filters

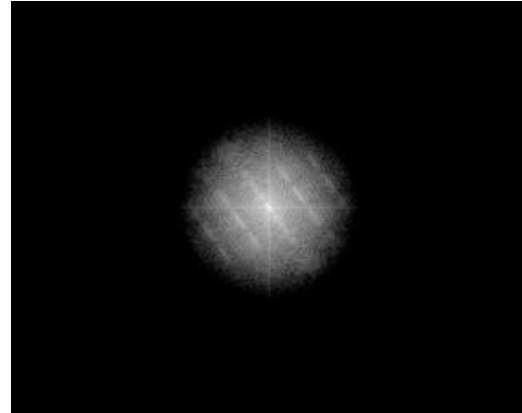
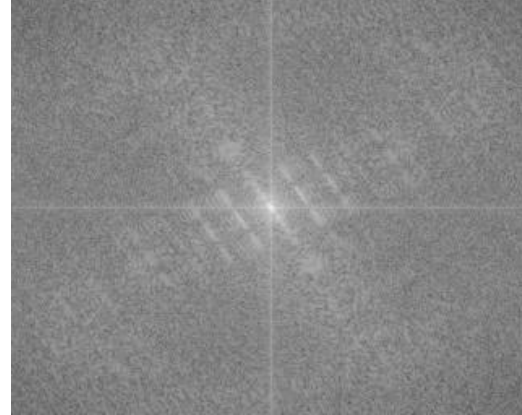
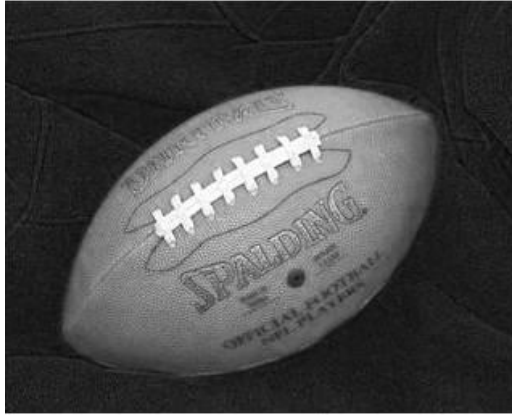
$$H(u, v) = \begin{cases} 1 & \text{if } D(u, v) \leq D_0 \\ 0 & \text{if } D(u, v) > D_0 \end{cases}$$



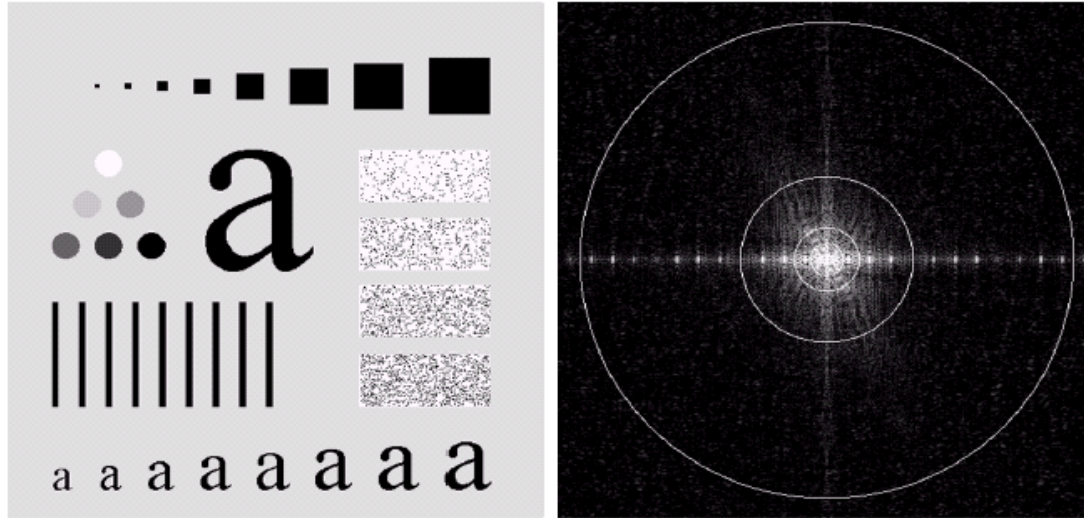
where $D(u, v) = [(u - M / 2)^2 + (v - N / 2)^2]^{1/2}$

$D_0 \rightarrow$ cut off frequency

Ideal Low Pass Filters

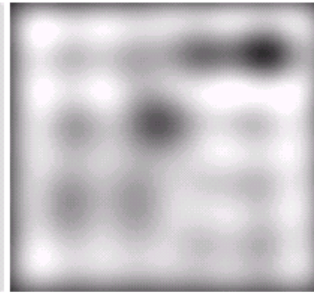
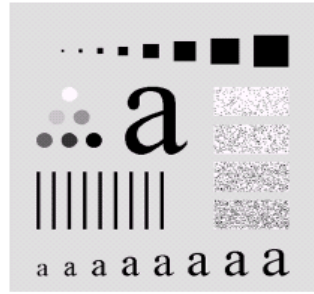


Ideal Low Pass Filters

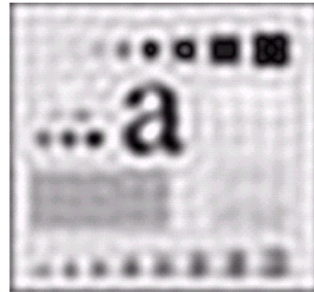


Radii 10,30,60,160 and 460 \rightarrow power 87, 93.1, 95.7, 97.8 and 99..2

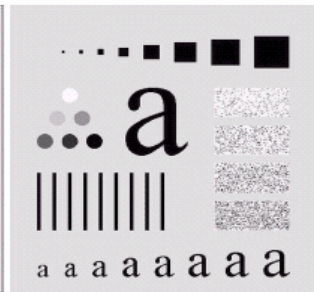
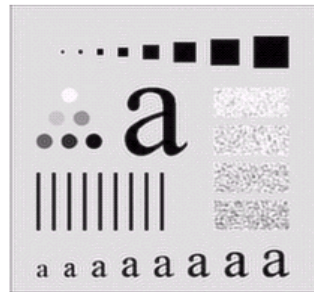
Ideal Low Pass Filters



ILPF radius 10



ILPF radius 60

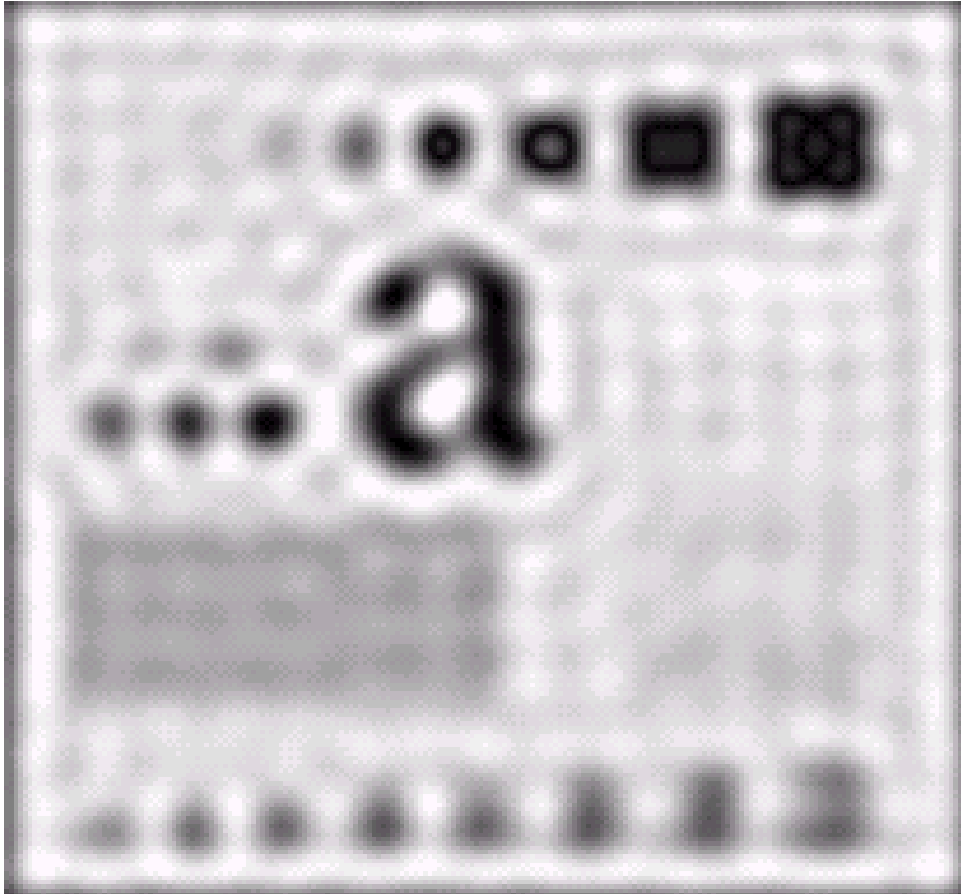


ILPF radius 460

ILPF radius 30

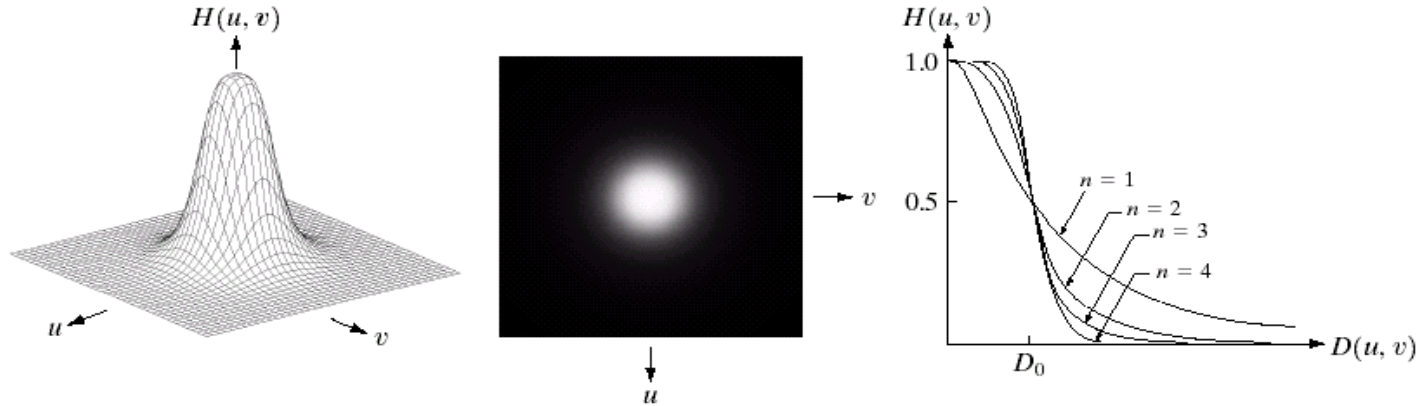
ILPF radius 160

Ideal Low Pass Filters



ILPF radius 30

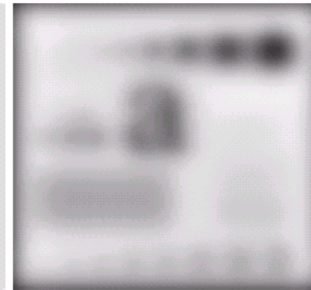
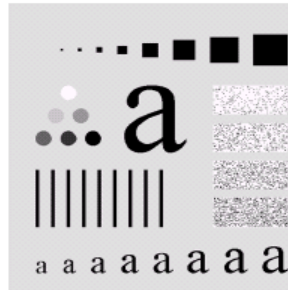
Butterworth Low Pass Filters



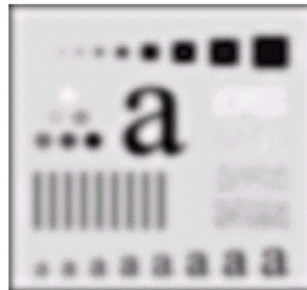
$$H(u, v) = \frac{1}{1 + [D(u, v) / D_0]^{2n}} \quad \text{where} \quad D(u, v) = [(u - M / 2)^2 + (v - N / 2)^2]^{1/2}$$

Butterworth Low Pass Filters (BLPF)

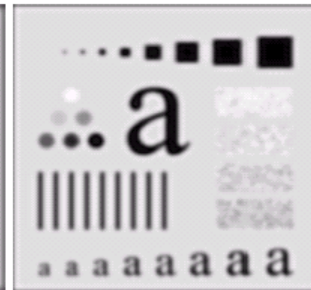
Order two, i.e.
 $n=2$



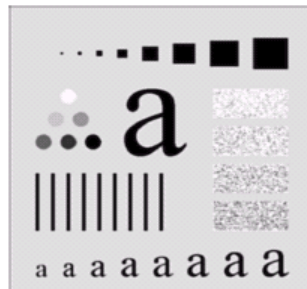
BLPF cut off
frequency 10



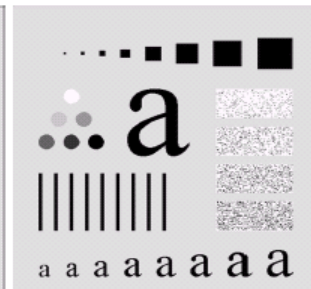
BLPF cut off
frequency 30



BLPF cut off
frequency 60

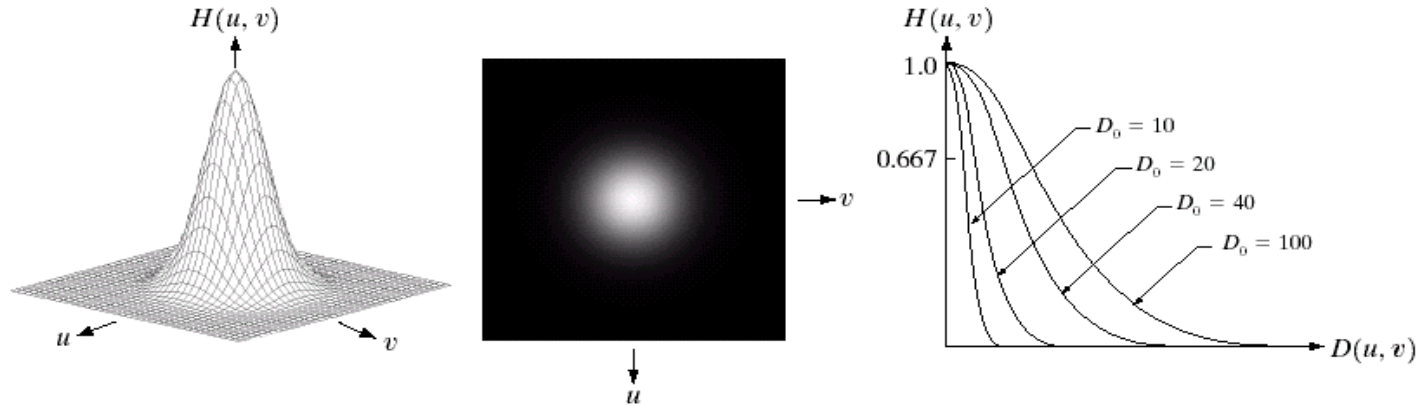


BLPF cut off
frequency 160



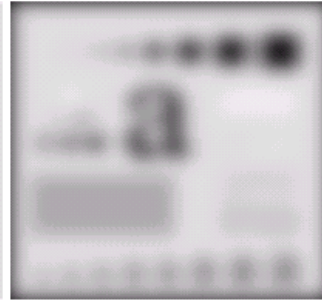
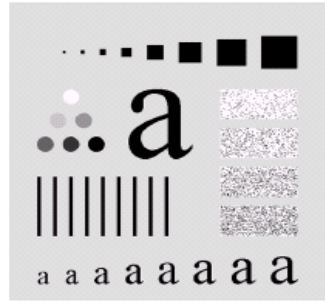
BLPF cut off
frequency 460

Gaussian Low Pass Filters

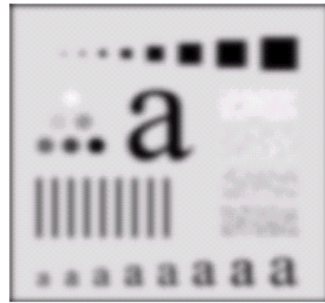


$$H(u, v) = e^{-D^2(u, v) / 2D_0^2}$$

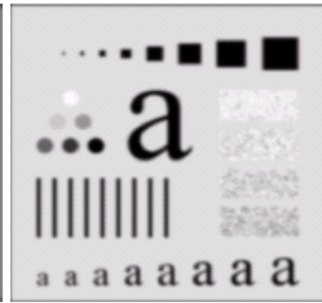
Gaussian Low Pass Filters (GLPF)



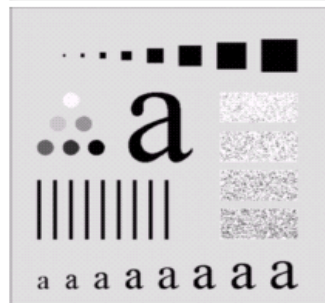
GLPF cut off
frequency 10



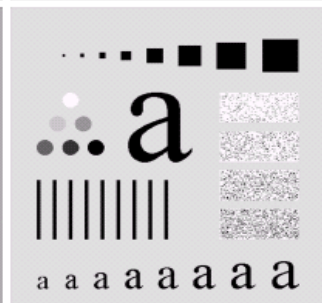
GLPF cut off
frequency 30



GLPF cut off
frequency 60



GLPF cut off
frequency 160



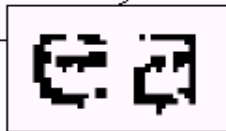
GLPF cut off
frequency 460

Comparison (ILPF, BLPF, GLPF)



Low pass filtering application

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.

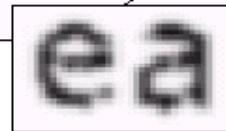
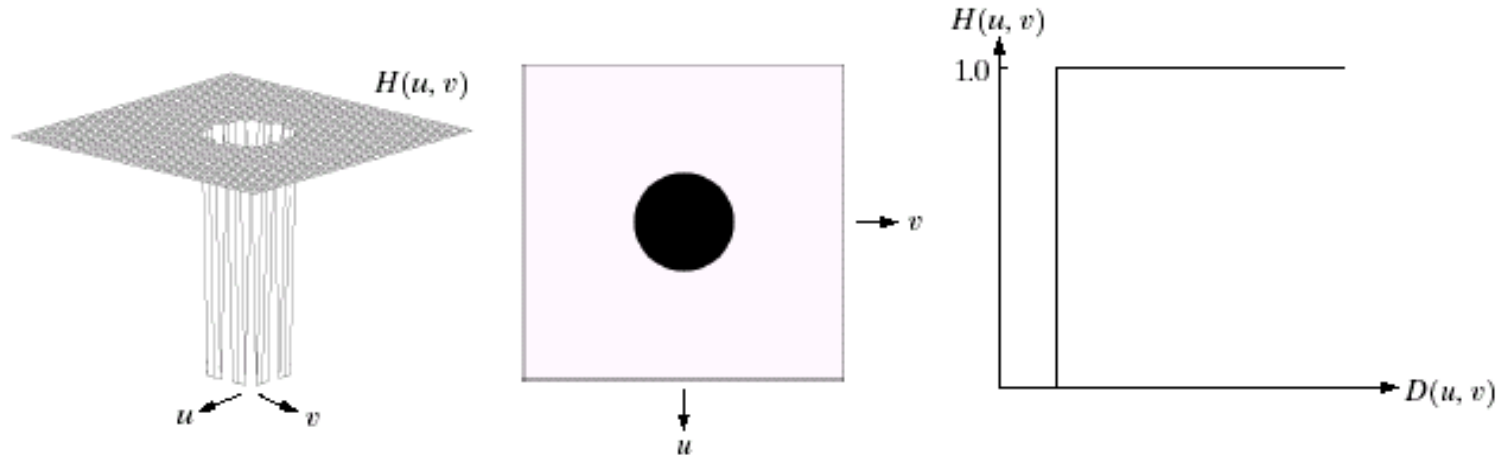


Image Sharpening in Frequency Domain

High Pass filter can be obtained from a given low pass filter:

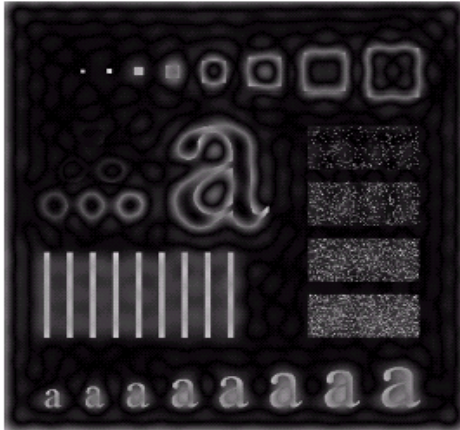
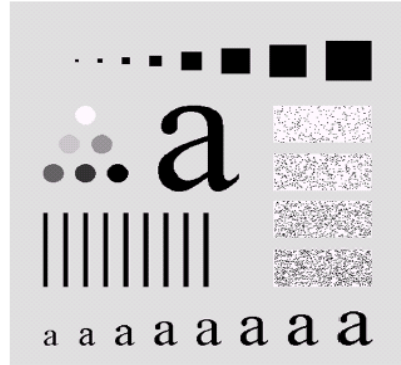
$$H_{hp}(u, v) = 1 - H_{lp}(u, v)$$

Ideal High Pass Filters



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

Ideal High Pass Filters



IHPL with $D_0 = 30$

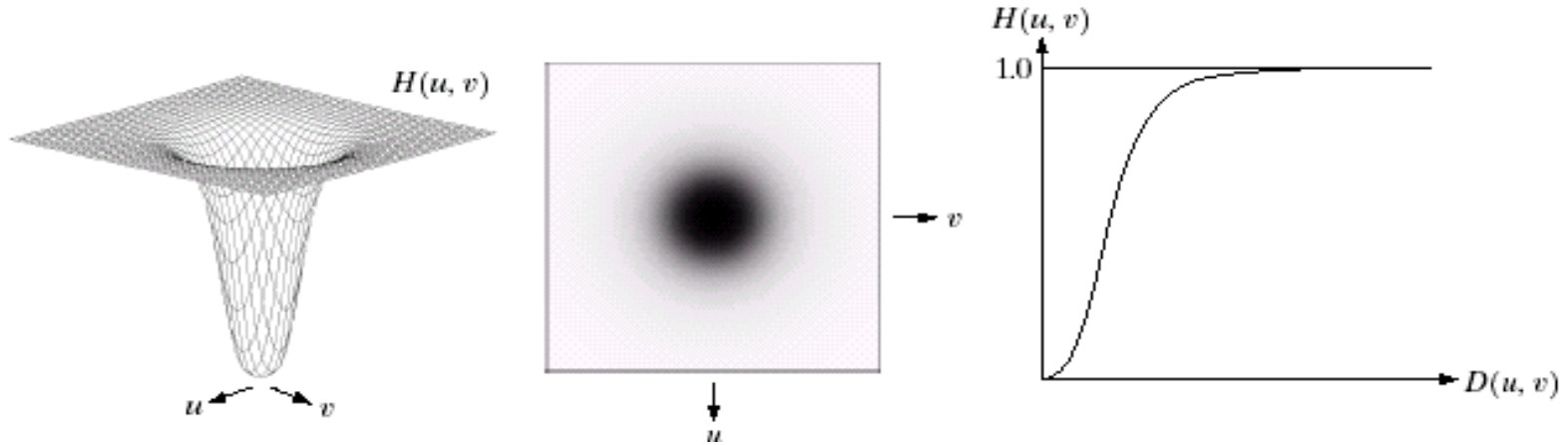


IHPF with $D_0 = 60$



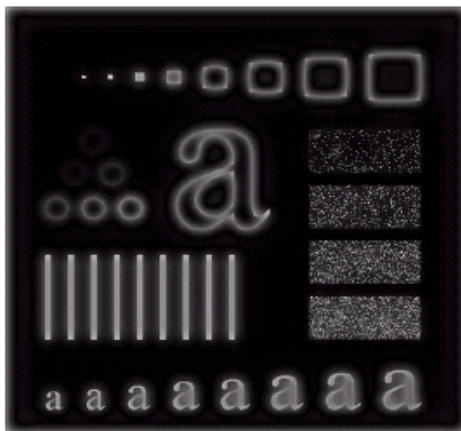
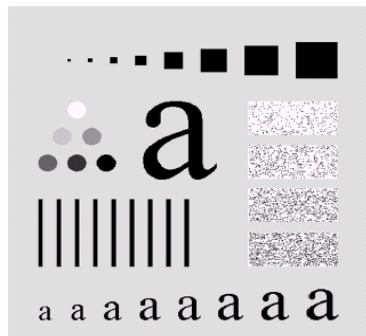
IHPF with $D_0 = 160$

Butterworth High Pass Filters

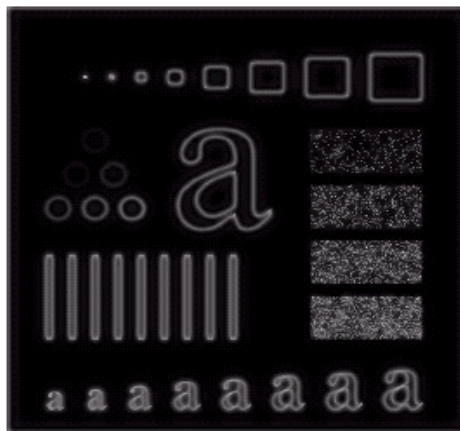


$$H(u, v) = \frac{1}{1 + [D_0 / D(u, v)]^{2n}}$$

Butterworth High Pass Filters



BHPL with $D_0 = 30$

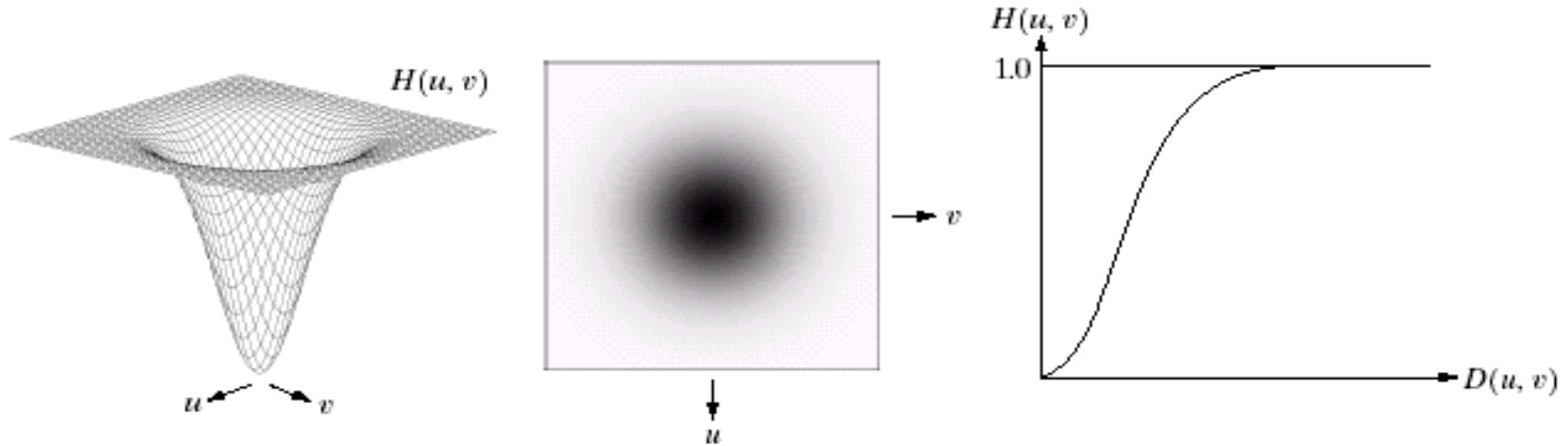


BHPF with $D_0 = 60$



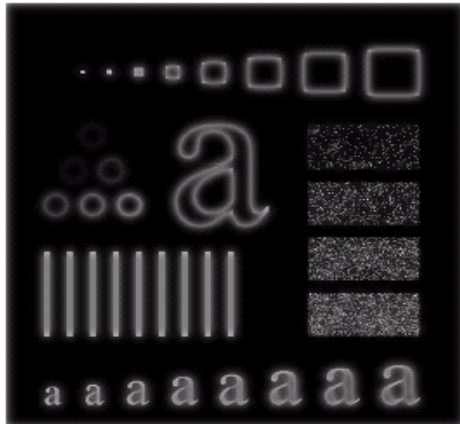
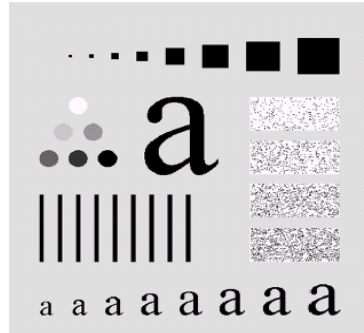
BHPF with $D_0 = 160$

Gaussian High Pass Filters



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

Gaussian High Pass Filters



GHPL with $D_0 = 30$



GHPF with $D_0 = 60$



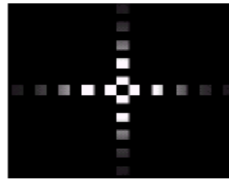
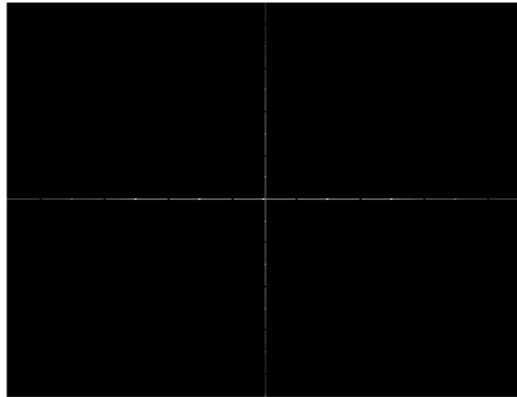
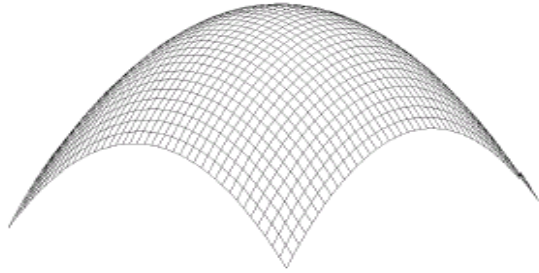
GHPF with $D_0 = 160$

Laplacian in frequency domain

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

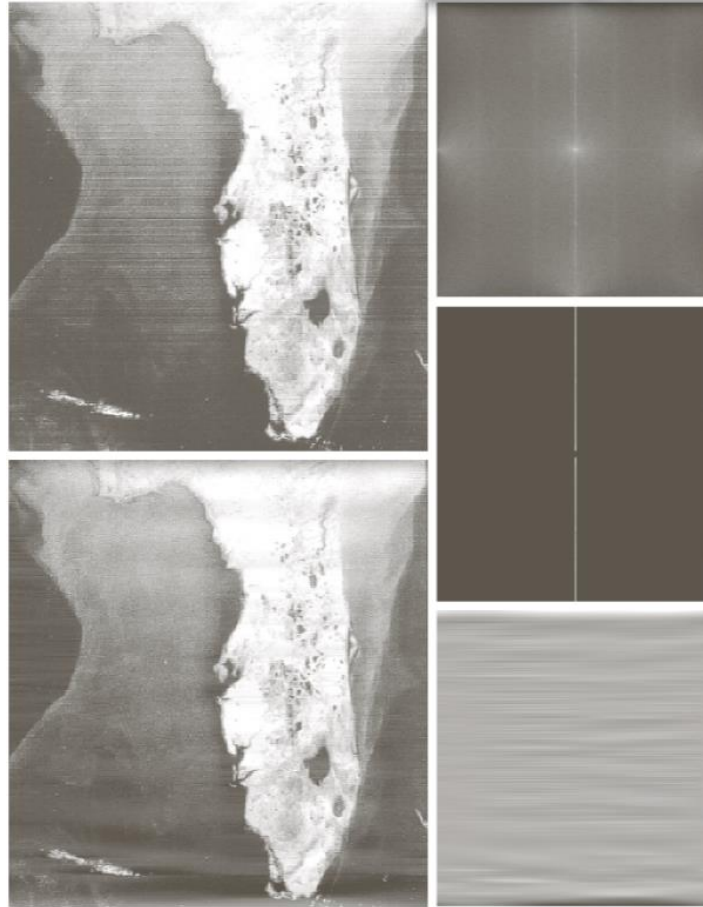
$$\begin{aligned}\mathfrak{F}\left[\frac{\partial^2(f(x, y))}{\partial x^2} + \frac{\partial^2(f(x, y))}{\partial y^2}\right] &= (ju)^2 F(u, v) + (jv)^2 F(u, v) \\ &= -(u^2 + v^2) F(u, v)\end{aligned}$$

Laplacian in frequency domain



0	1	0
1	-4	1
0	1	0

Notch Reject filter (Notch pass filter)



Filtering in frequency domain

- Band reject (Band pass filters)
- Unsharp Masking and High boost filtering
- Homomorphic filtering