



UNIVERSITEIT  
GENT

# Multimedia – Labo 3

**Gianni Allebosch, Martin Dimitrievski**  
**{gianni.allebosch, martin.dimitrievski}@ugent.be**

**Master of Science in de industriële wetenschappen: elektronica-ICT**

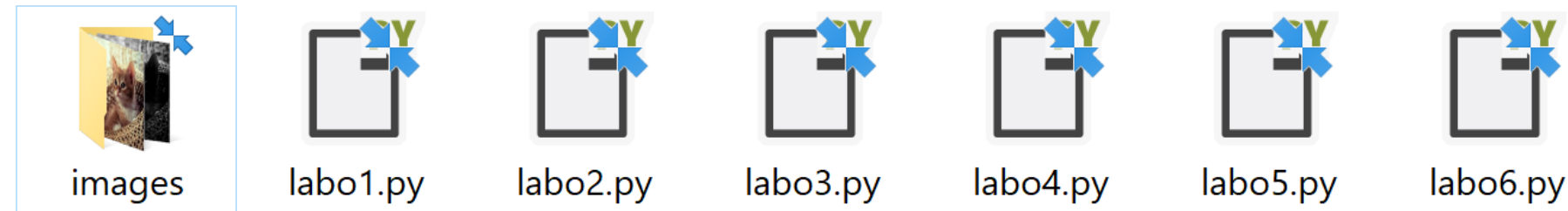
**Campus Ardoyen,  
BE - 9000 Gent**

# TIPS & TRICKS

- **Keep your program user friendly!**
  - Provide a README.txt
  - Execute each exercise sequentially in 1 script
  - Make clear which exercise is currently being executed
  - Name your windows when showing images
  - If input is required from the user, make sure the instructions are clear

# TIPS & TRICKS

- **Be sure the code is executable as is!**
  - We will not change your code!
  - Functions in comments will not be executed
  - Store all input images in a folder named “images” next to your code



- Do not use absolute paths but use relative paths (portability!):

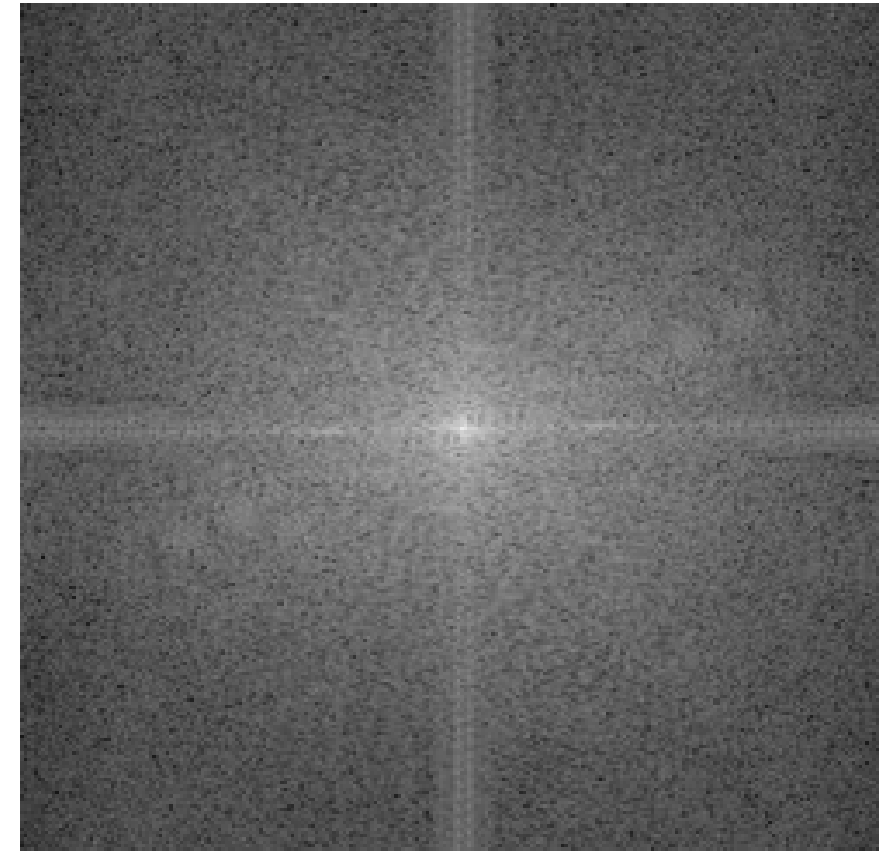
✗ “C:/users/admin/multimedia/labo1/images/lena\_color.jpg”

✓ “images/lena\_color.jpg”

# 2D Discrete Fourier Transform



image



fft magnitude spectrum

Frequency domain  $\rightarrow$  spatial changes/variations in intensity

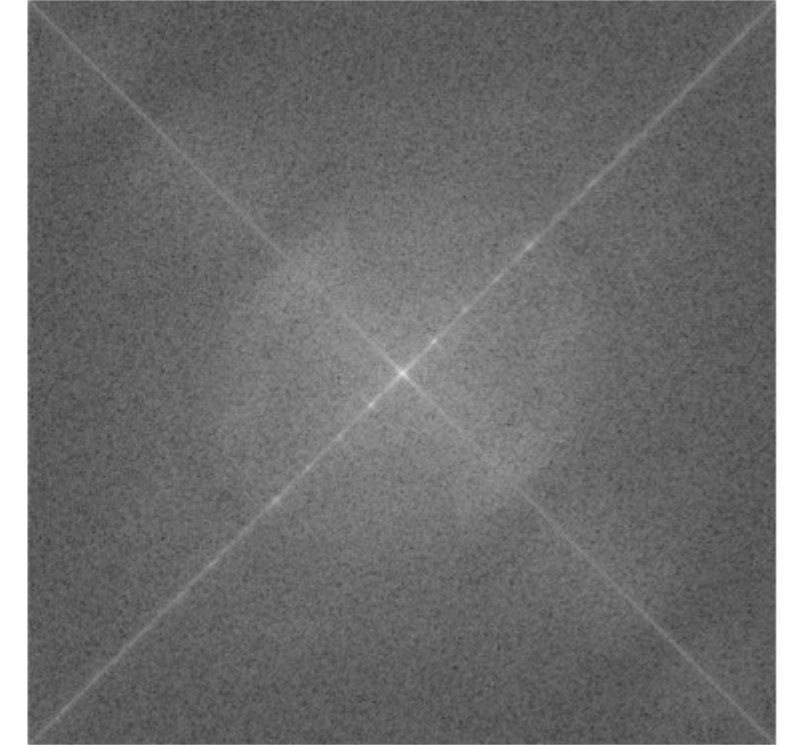
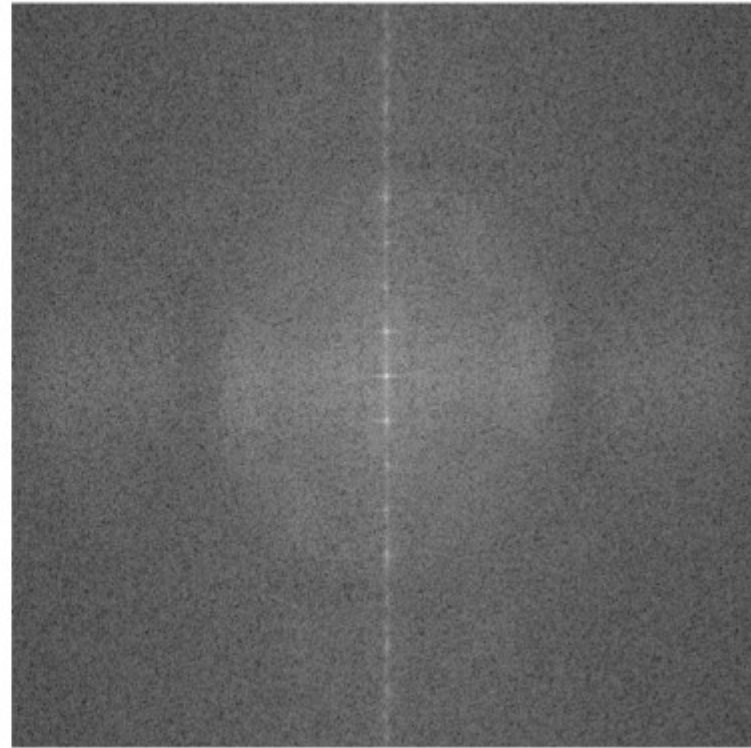
- Low frequencies (close to the origin) = slowly changing variations in image e.g. a plane of homogeneous color
- High frequencies (away from the origin) = sudden changes in the image e.g. edges or noise
- DC-component (the origin) = average gray value of the image

# 2D Discrete Fourier Transform

## Sonnet for Lena

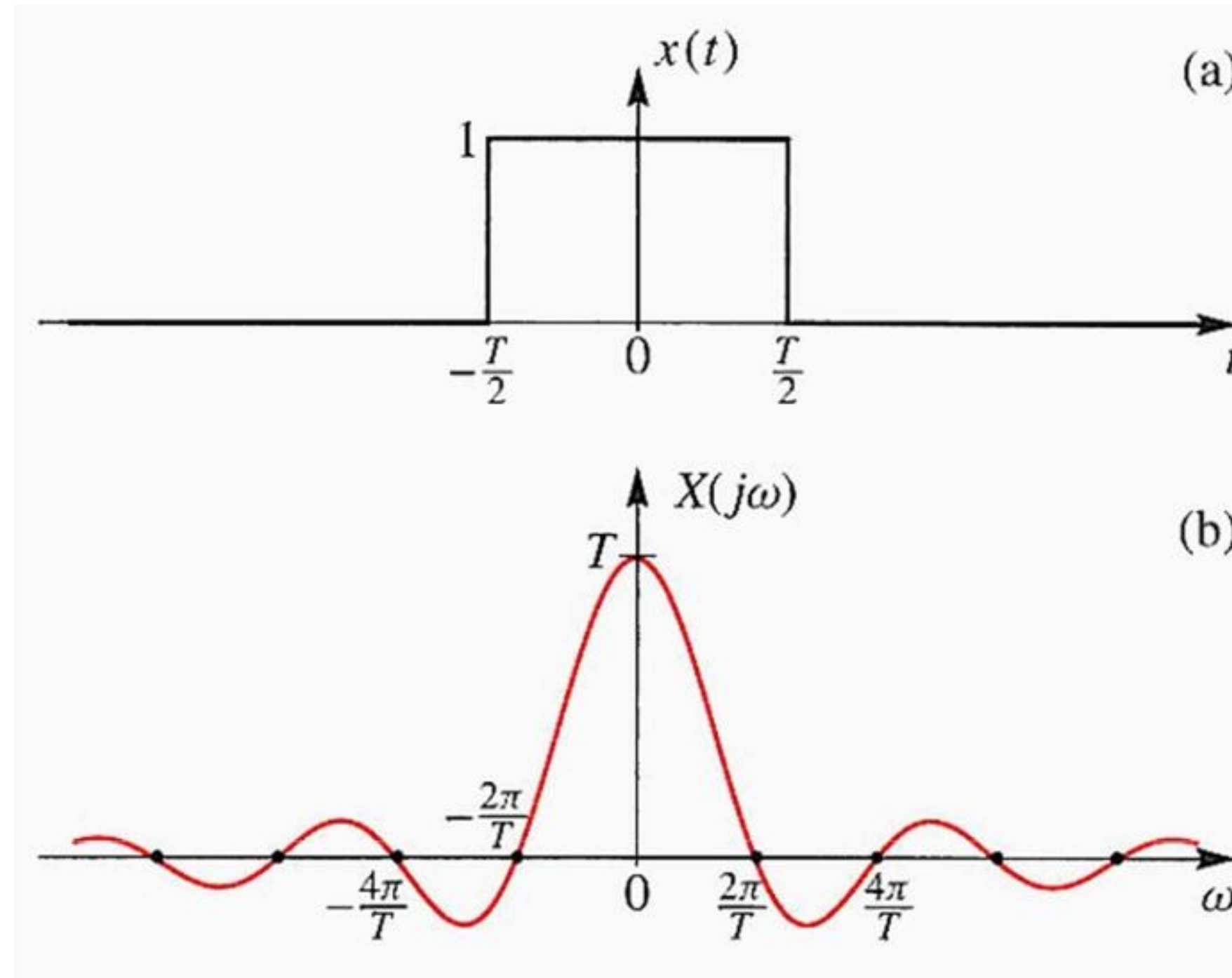
O dear Lena, your beauty is no want  
It is hard sometimes to describe it fast.  
I thought the entire world I would impress  
If only your portrait I could compress.  
Alas! First when I tried to use VQ  
I found that your cheeks belong to only you.  
Your silky hair contains a thousand lines  
Hard to match with sums of discrete cosines.  
And for your lips, sensual and tactual  
Thirteen Crays found not the proper fractal.  
And while these setbacks are all quite severe  
I might have fixed them with hacks here or there  
But when filters took sparkle from your eyes  
I said, 'Damn all this. I'll just digitize.'

Thomas Goldhurst



Frequency domain → spatial changes/variations in intensity  
→ Major directions in images

# 2D Discrete Fourier Transform



## Time domain

Rectangular pulse with

$T$  = pulse width

$A$  = pulse amplitude

## Fourier Transform

## Frequency domain

Fading sinus (= ? Function) with

$1/T$  = frequency

$2\pi/T$  = angular frequency



# Question – 2D Discrete Fourier Transform

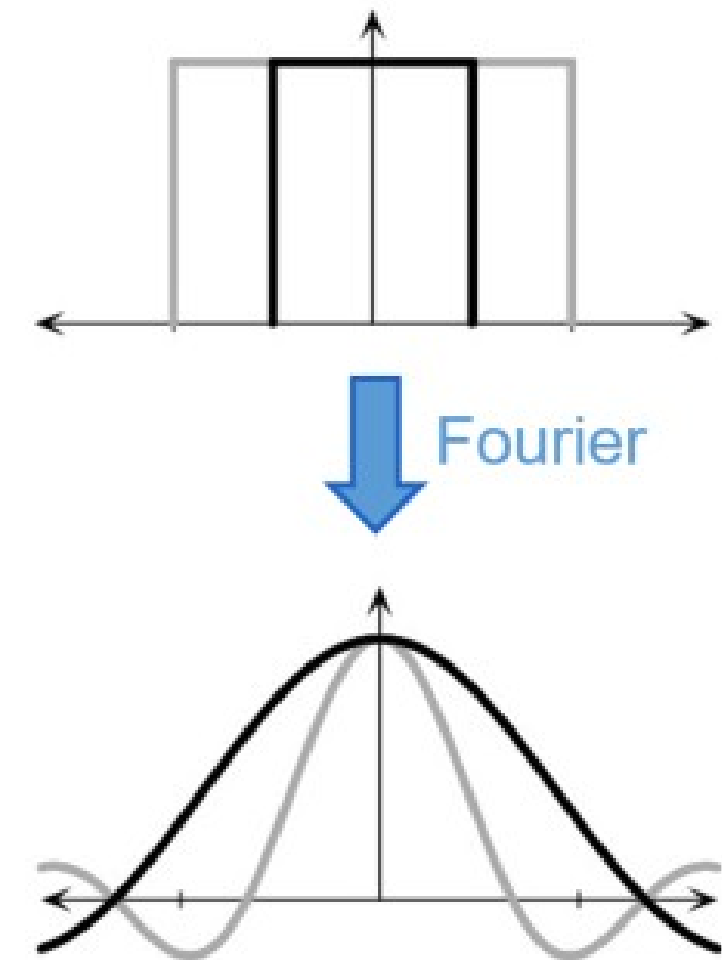
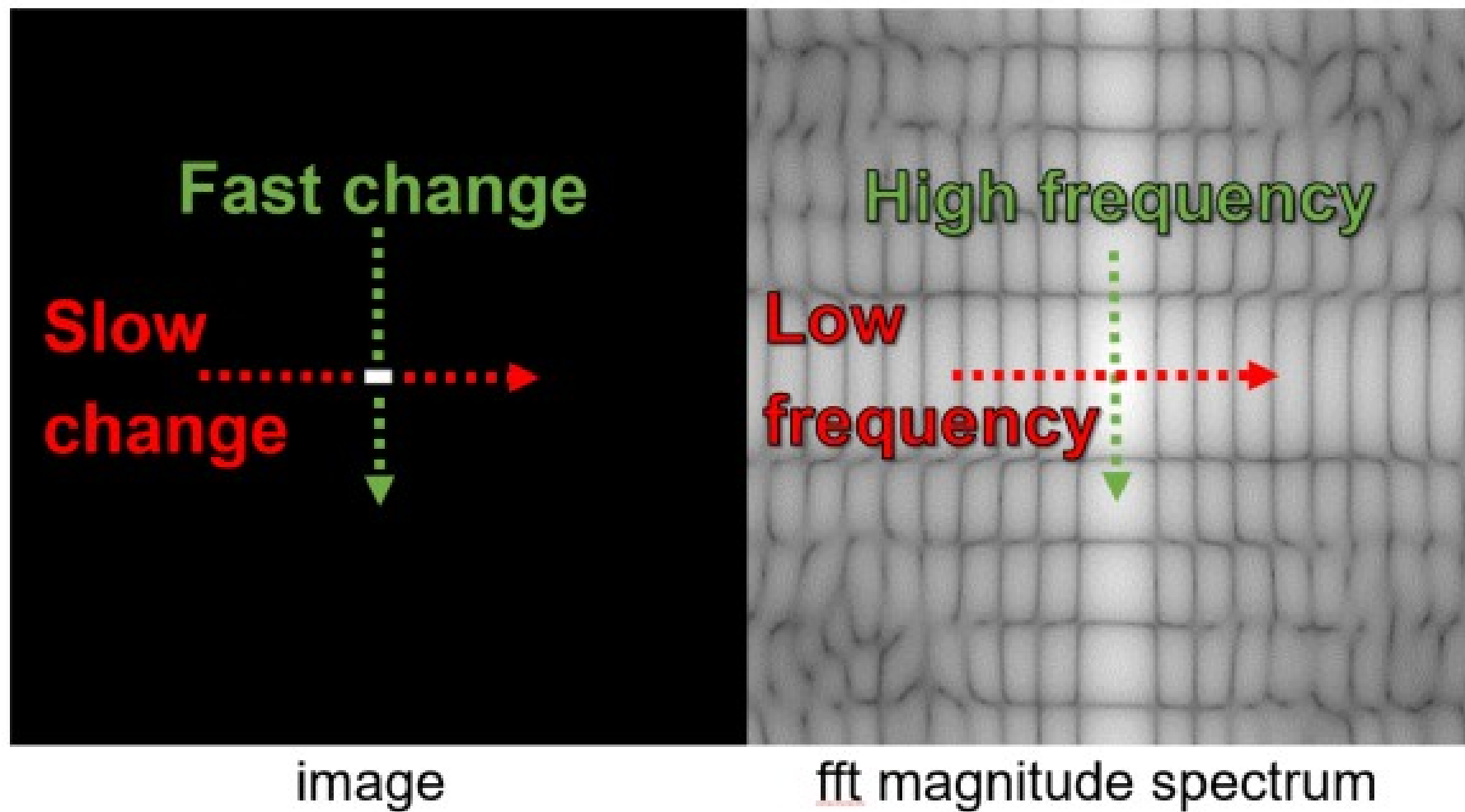
**Solve the question in your README.txt**

According to what pattern are the zero points in the spectrum formed?  
Why?



# 2D Discrete Fourier Transform

2D Fourier  $\rightarrow$  Fourier in x and y direction



# 2D Discrete Fourier Transform

```
import numpy as np
import cv2
import matplotlib.pyplot as plt

image = cv2.imread("disk.jpg", cv2.IMREAD_GRAYSCALE) # grayscale images
fbeeld = np.fft.fft2(image) # 2D Fourier transform
fshift = np.fft.fftshift(fbeeld) # shift origin to center
plt.imshow(magnitude_spectrum(fshift), cmap='gray') # show magnitude spectrum of fshift
plt.show()

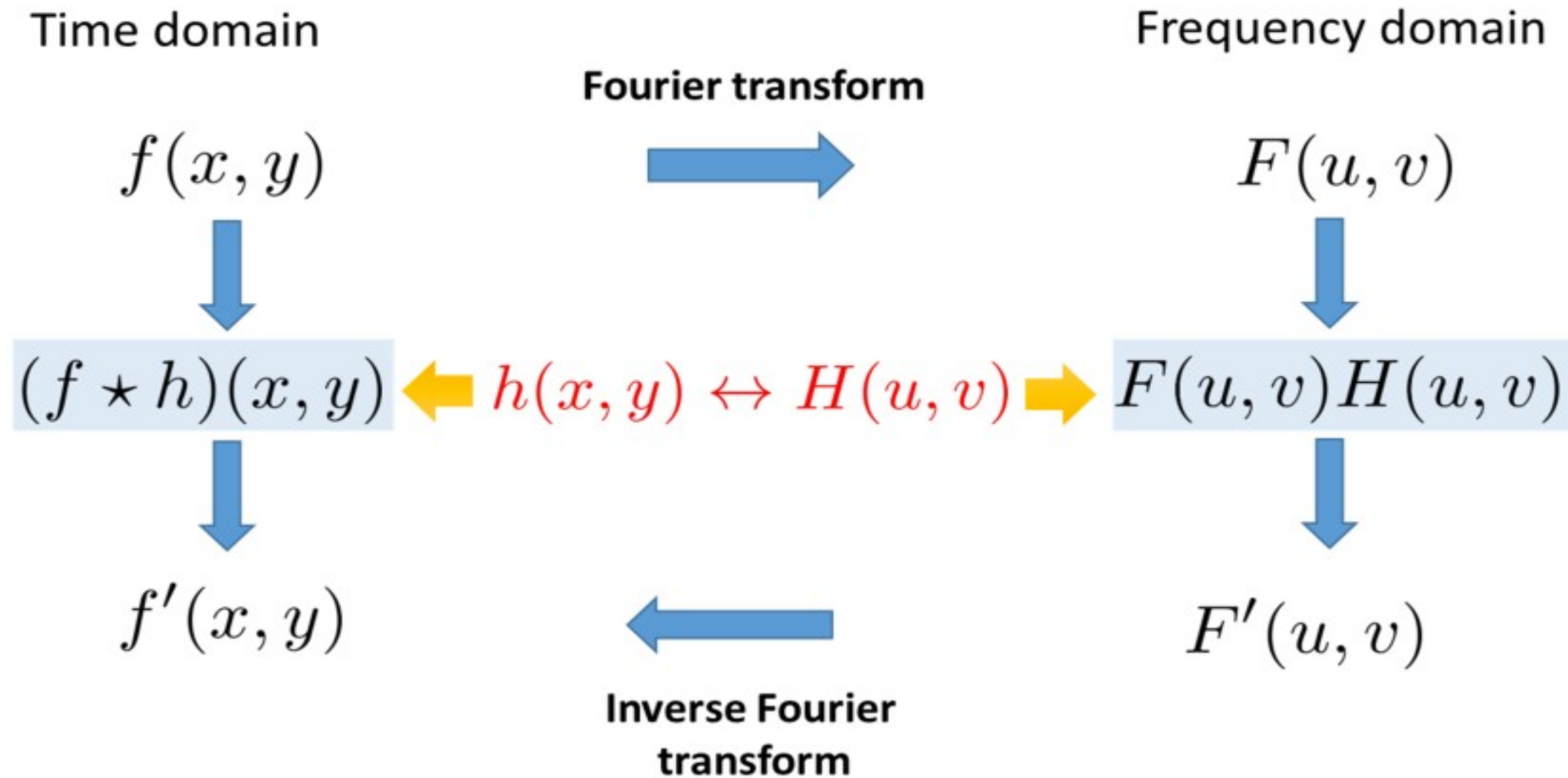
ifshift = np.fft.ifftshift(fshift) # shift origin to original place (is the same as fftshift but more readable)
ifbeeld = np.fft.ifft2(ifshift) # inverse 2D Fourier transform
ifbeeld = ifbeeld.real # complex (because of rounding errors) to real values
plt.imshow(ifbeeld, cmap='gray')
plt.show()

def magnitude_spectrum(fshift):
    """ Returns magnitude spectrum of the 2D fourier transform for viewing purpose. """
    magnitude_spectrum = np.log(np.abs(fshift)+1) # reduce large dynamic range
    magnitude_spectrum -= np.min(magnitude_spectrum) # scale to range [0, 255]
    magnitude_spectrum *= 255./np.max(magnitude_spectrum) # scale to range [0, 255]
    return magnitude_spectrum
```

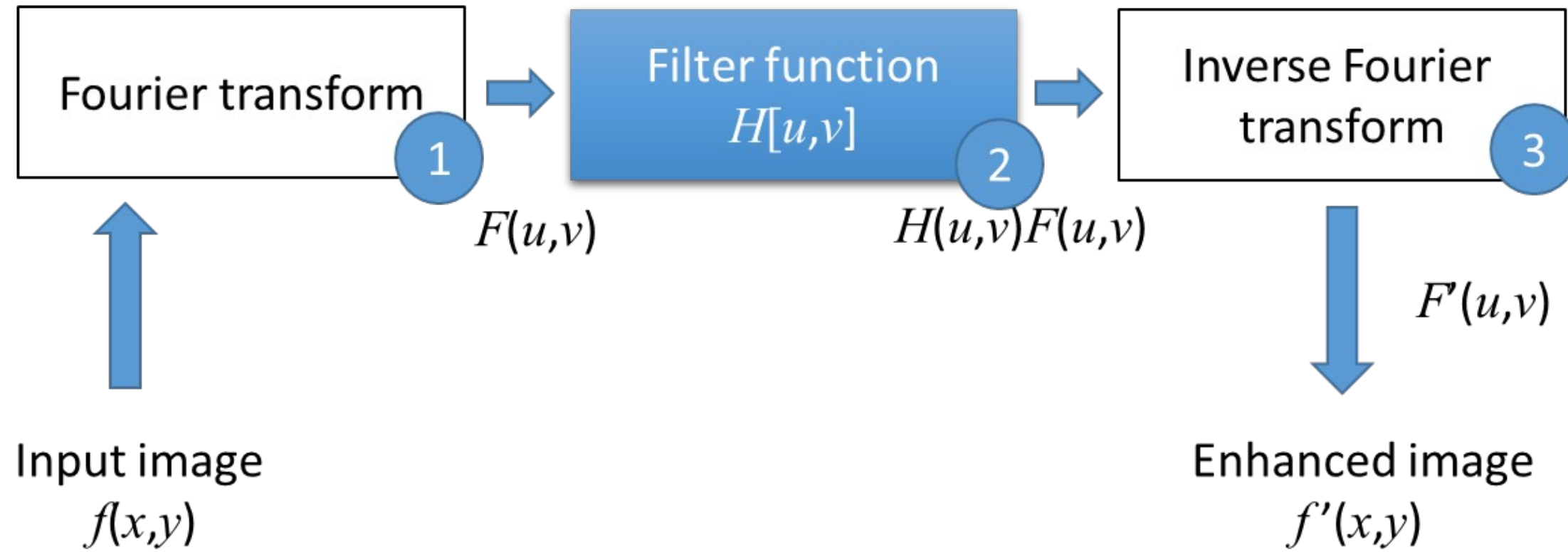
# Filtering in the frequency domain

Two equivalent ways of filtering:

Two equivalent ways of filtering



# Filtering in the frequency domain



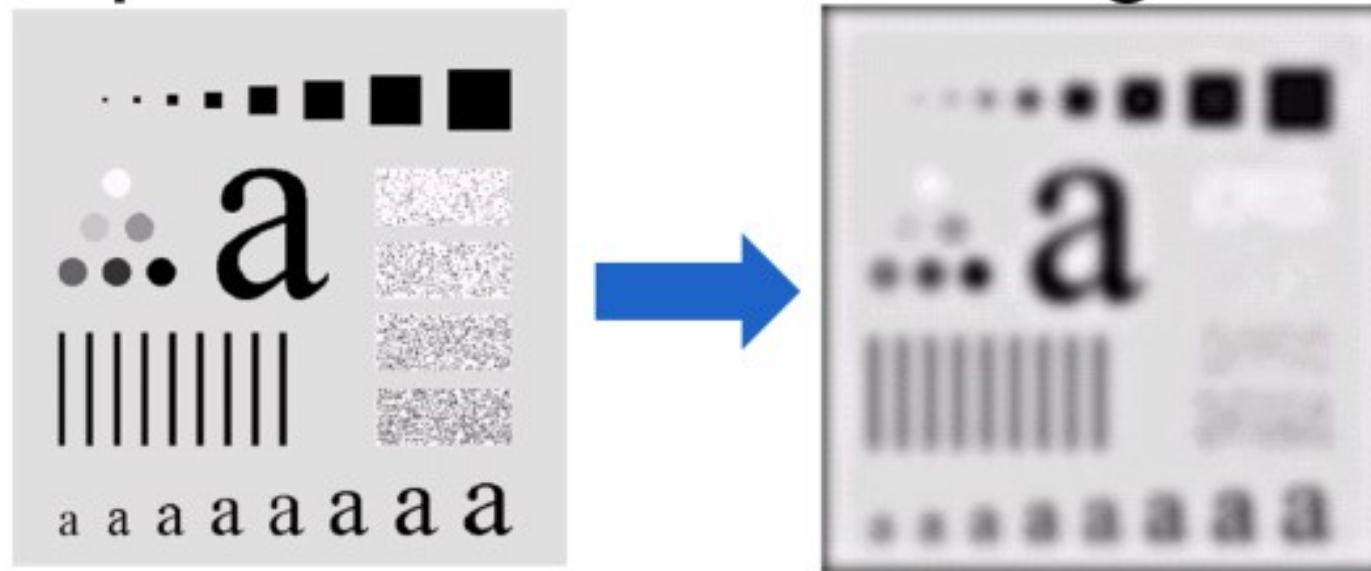
- 1  $F(u, v) = \mathcal{F} [f] (u, v)$
- 2  $F'(u, v) = H(u, v)F(u, v)$
- 3  $f'(x, y) = \mathcal{F}^{-1} (F') (x, y)$

Notation conventions:

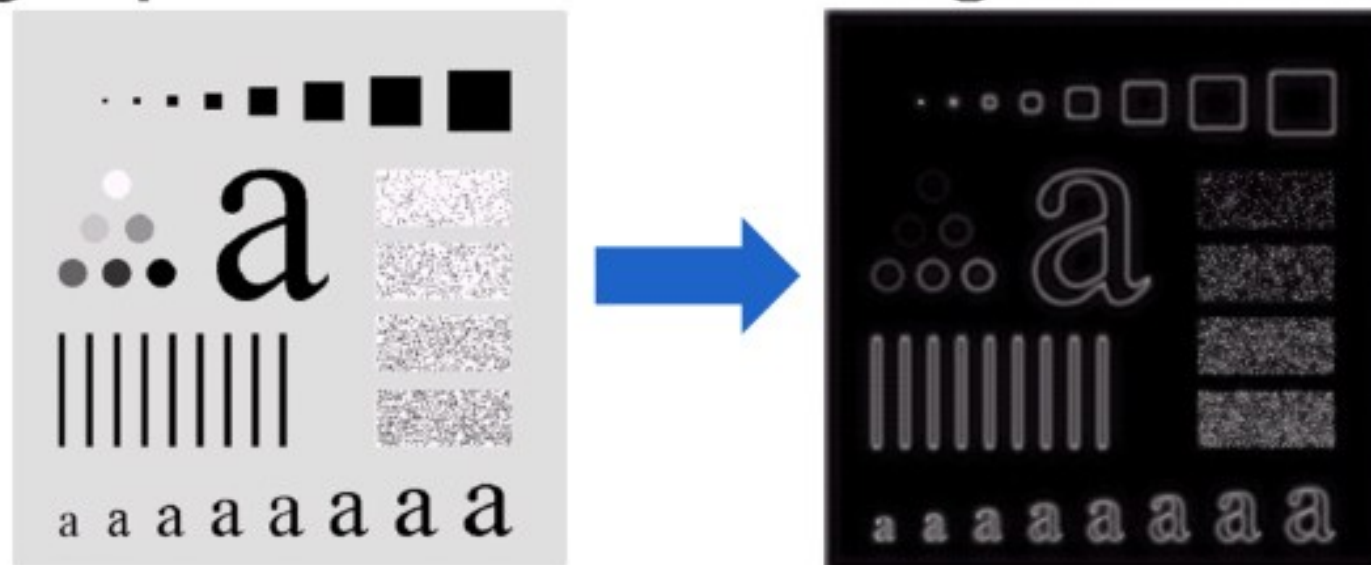
- Small letter: spatial domain
- Capital letter: frequency domain

# Filtering in the frequency domain

- Low-pass filters → blurring



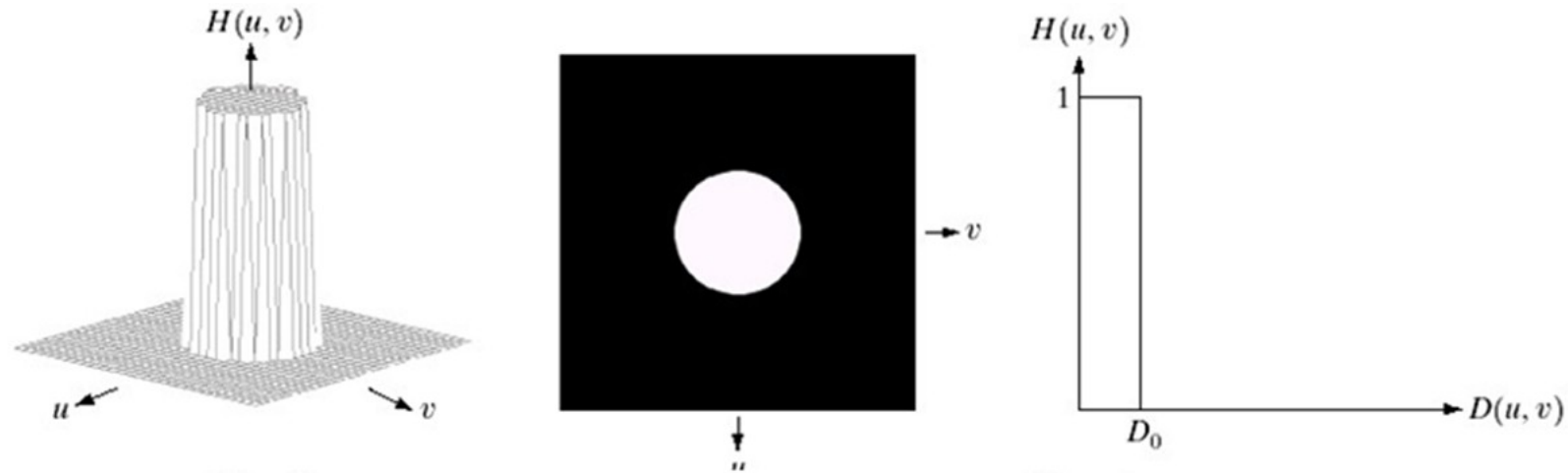
- High-pass filters → edge detection



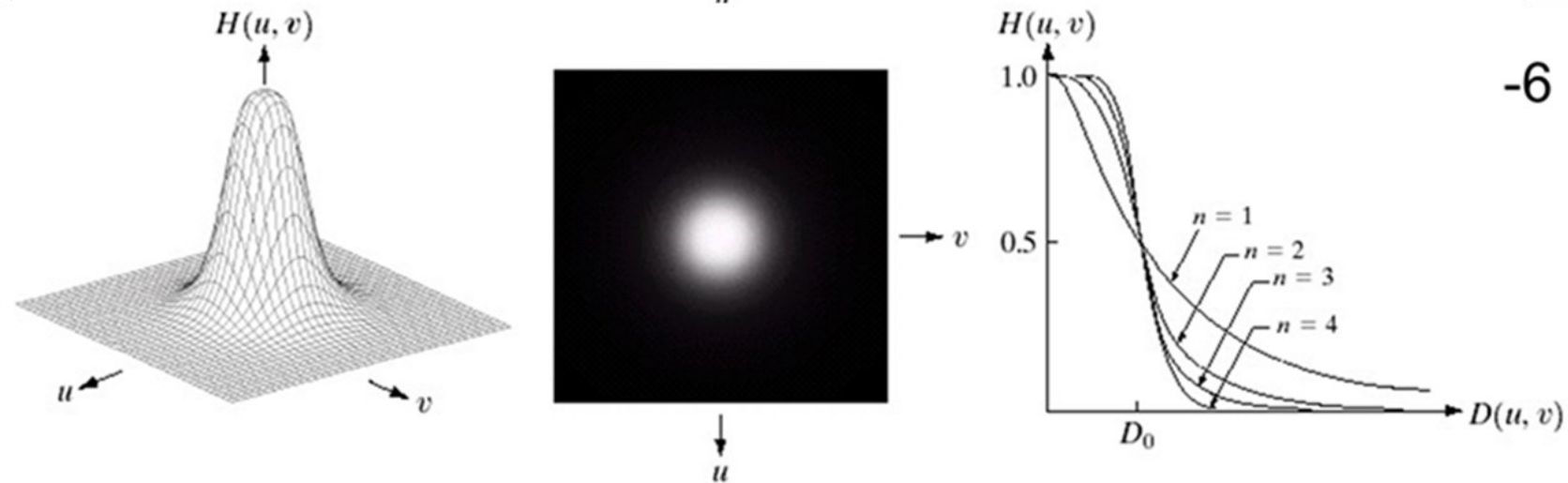


# Lowpass Filters

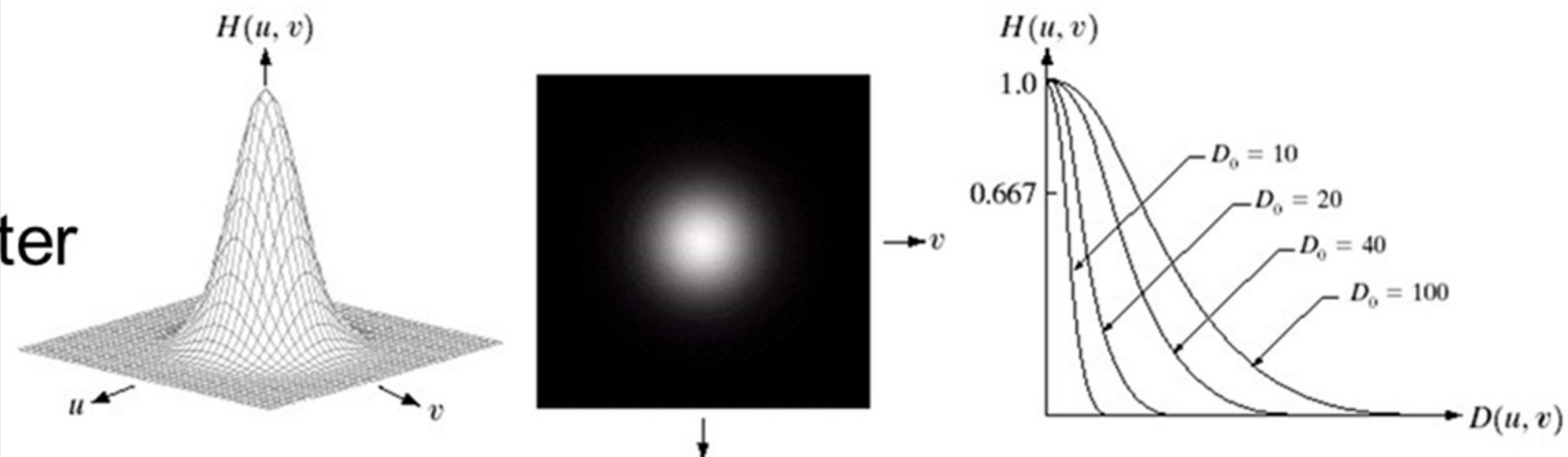
Ideal lowpass filter



Butterworth lowpass



Gaussian lowpass filter



a b c  
d e f  
g h i

**FIGURE 4.22** Top row: Perspective plot, image representation, and cross section of a typical ideal lowpass filter. Middle and bottom rows: The same sequence for typical Butterworth and Gaussian lowpass filters.

# Exercise 1 – Ideal Lowpass Filter

**Solve the questions in your README.txt**

- What does the frequency content with highest magnitude represent?
- Discuss what happens when the radius of the circular filter varies.
- Discuss the effects of the ideal lowpass filter and ringing.

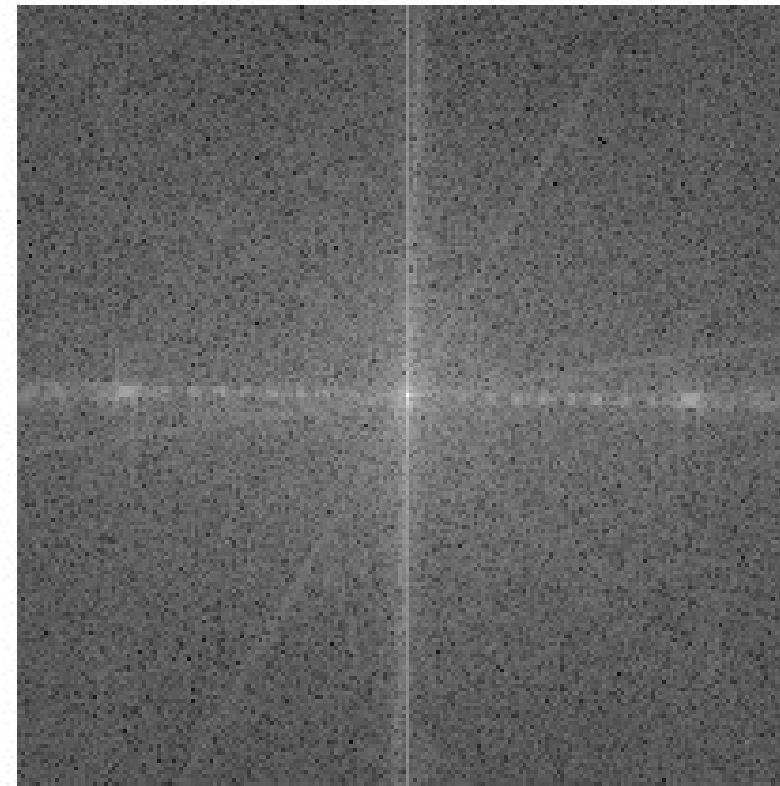


# Exercise 1 – Ideal Lowpass Filter

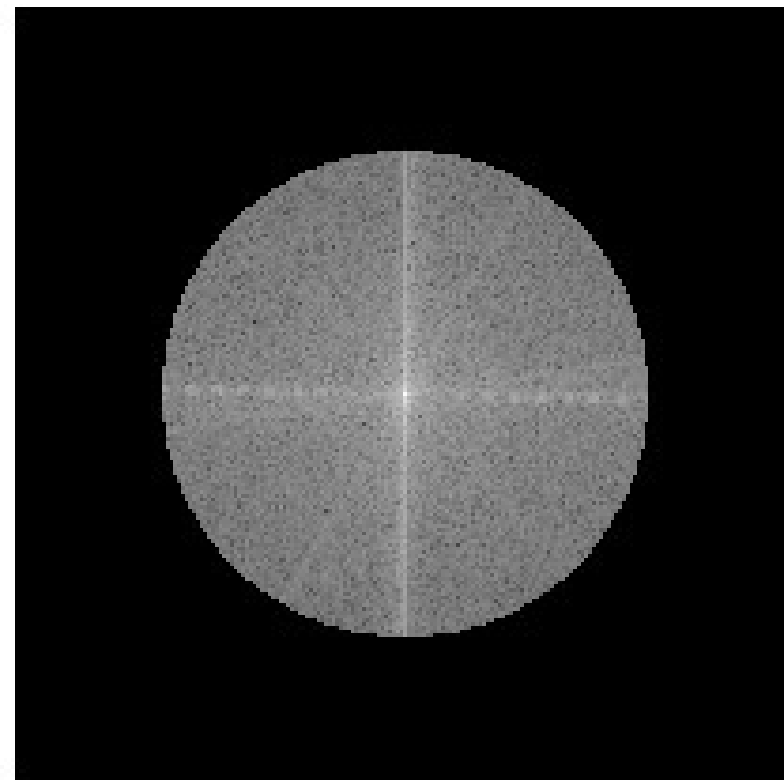
Original



FFT



Filtered FFT



Filtered Image

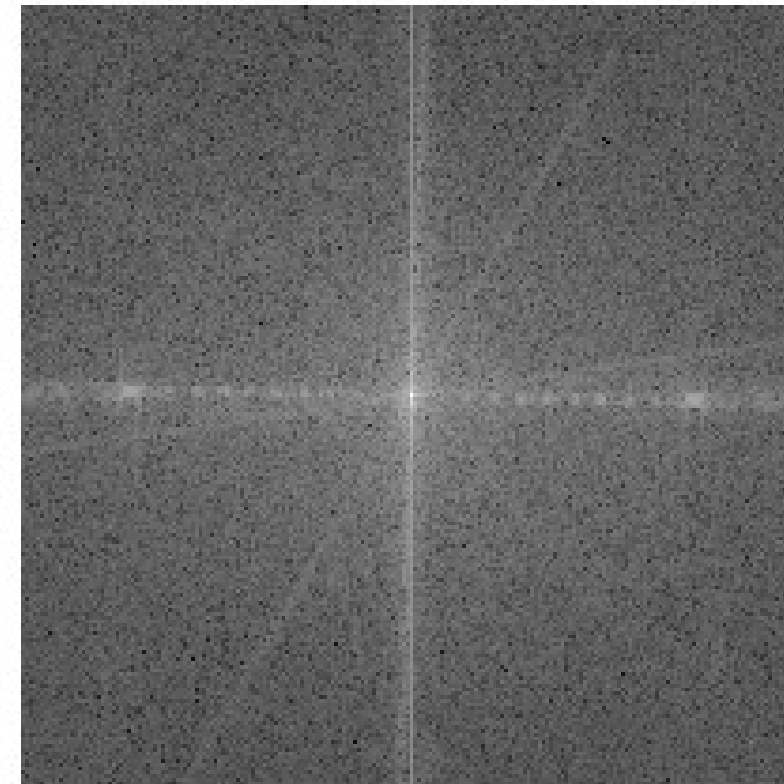


# Exercise 1 – Ideal Lowpass Filter

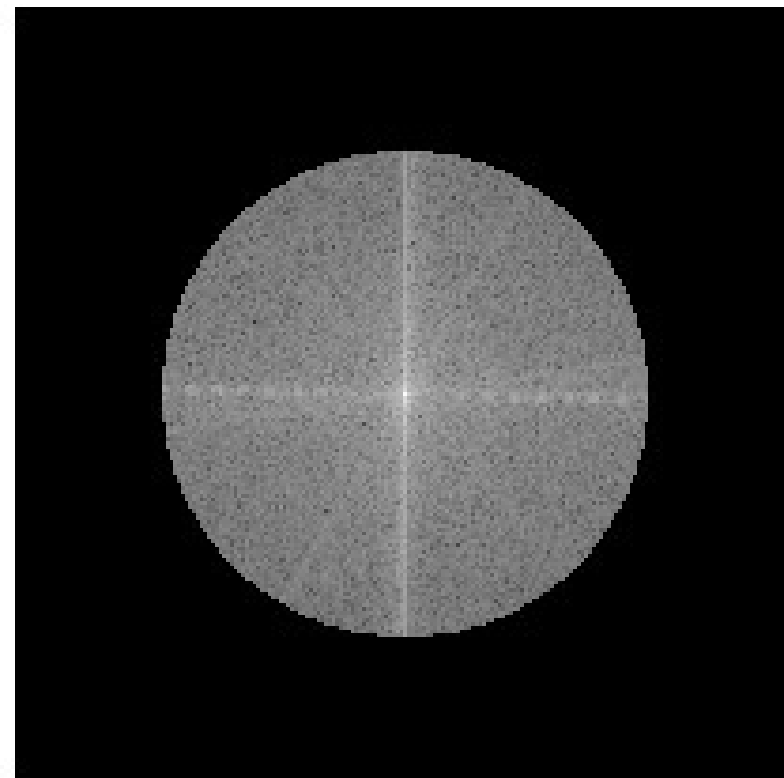
Original



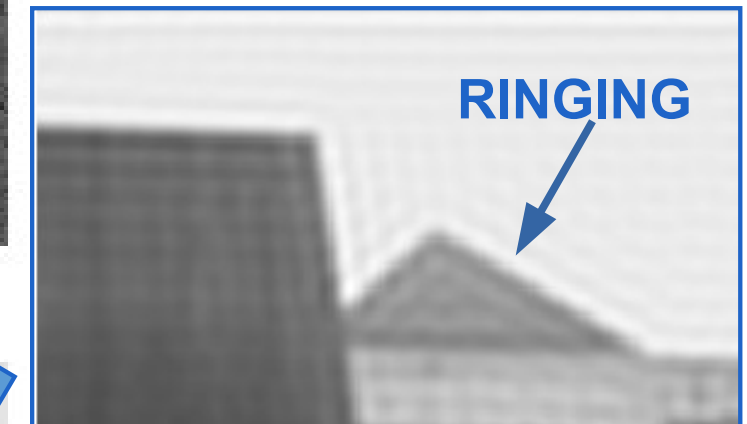
FFT



Filtered FFT

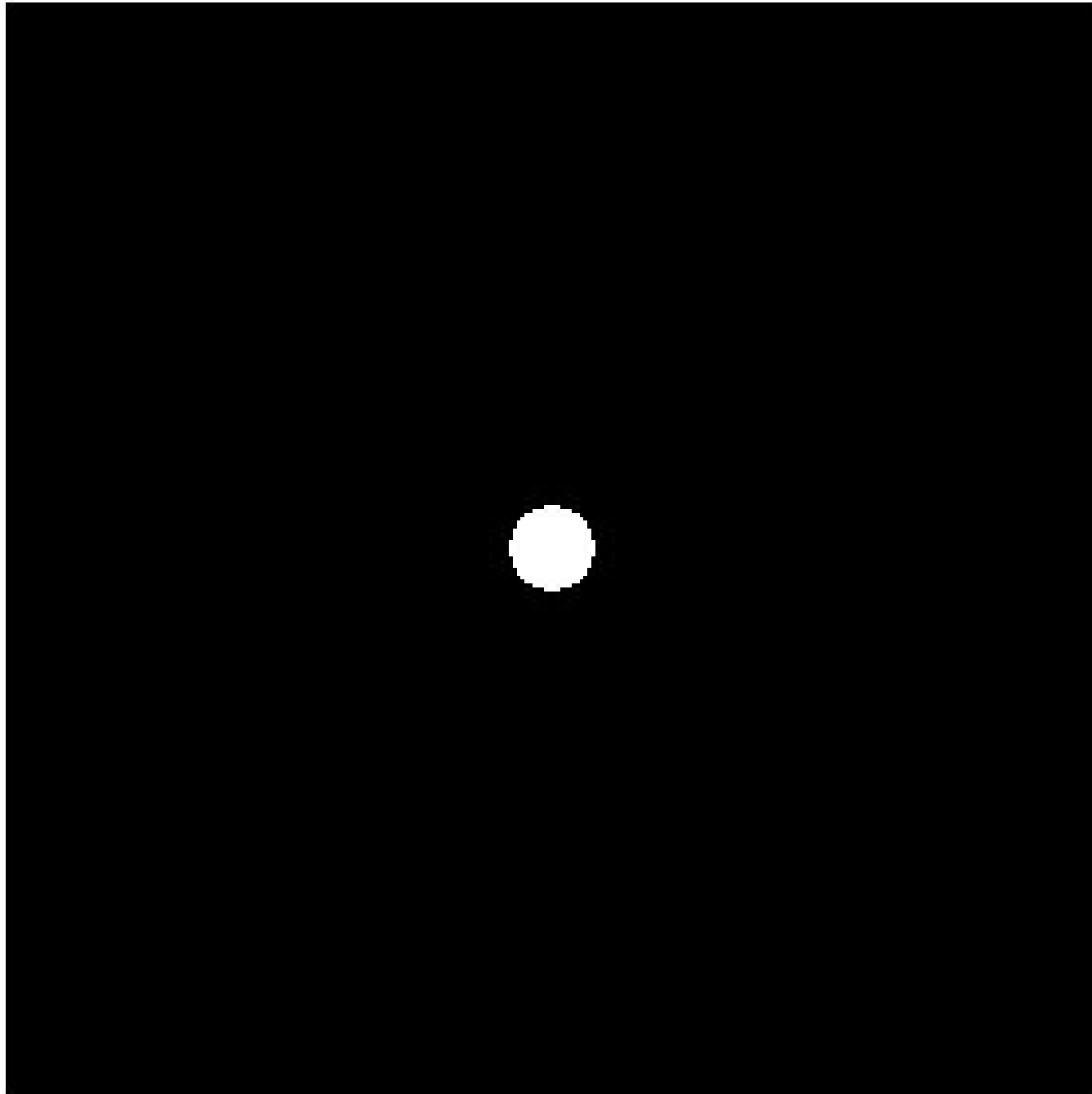


Filtered Image



# Exercise 1 – Ideal Lowpass Filter

Filter in frequency domain



Filter in spatial domain



# Exercise 2 – Gaussian Lowpass Filter

**Solve the questions in your README.txt**

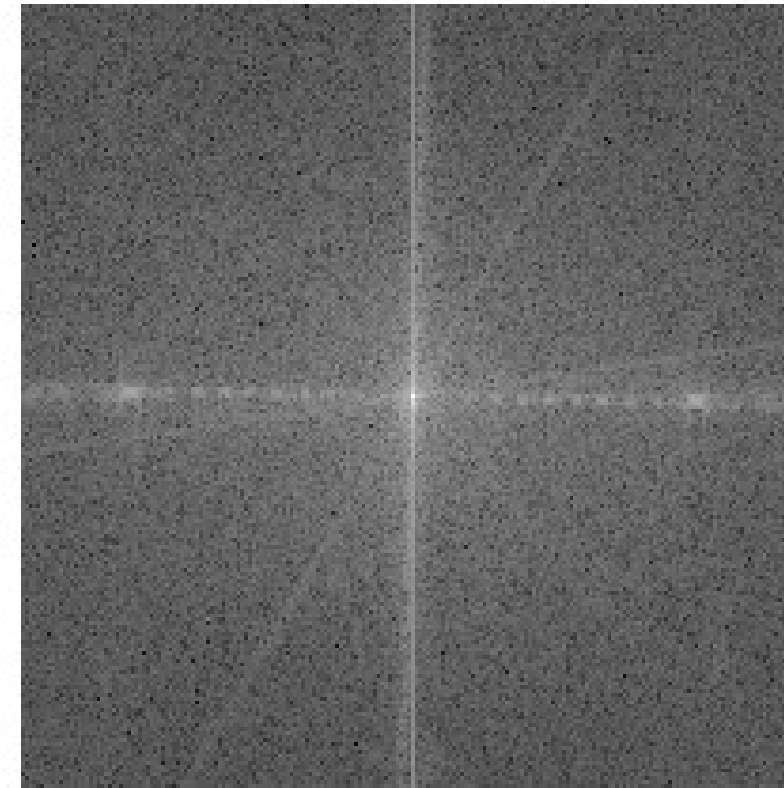
- Discuss the advantages and disadvantages of the Gaussian lowpass filter.
- Discuss what happens when  $D_0$  varies.

# Exercise 2 – Gaussian Lowpass Filter

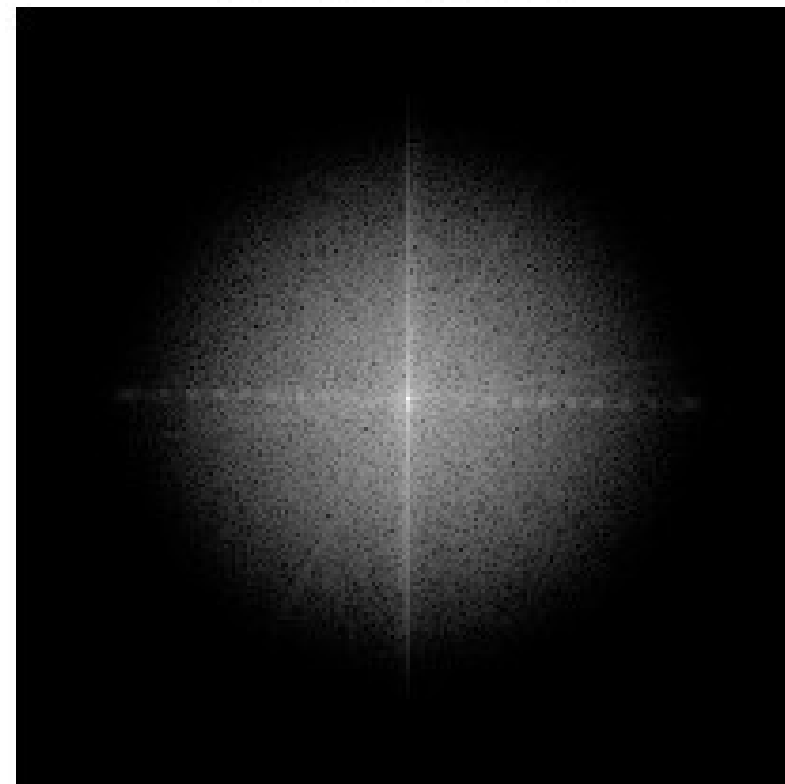
Original



FFT



Filtered FFT



Filtered Image



# Exercise 3 – Butterworth Lowpass Filter

**Solve the question in your README.txt**

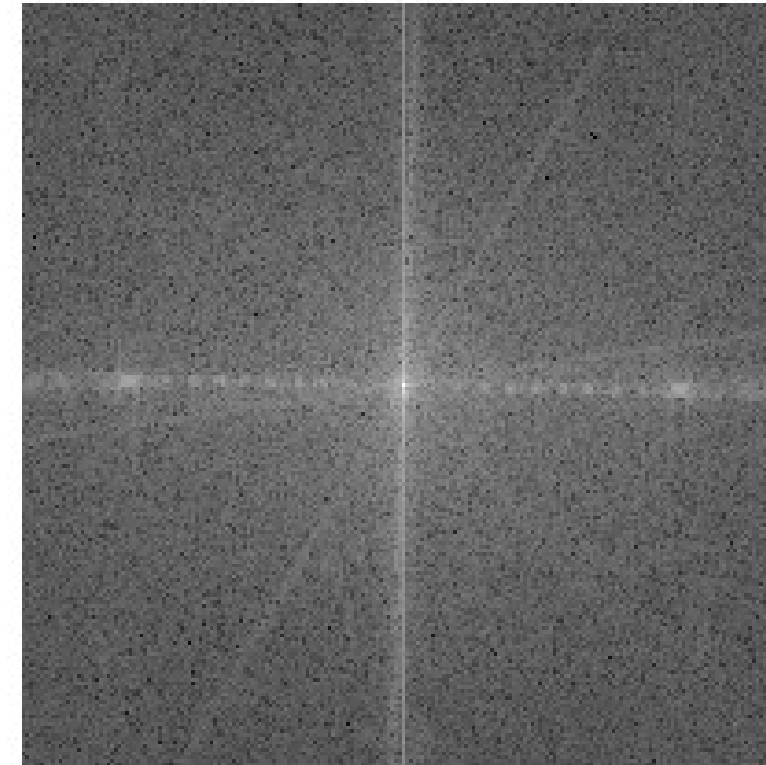
Discuss how the Butterworth lowpass filter solves some disadvantages of the Gaussian lowpass filter.

# Exercise 3 – Butterworth Lowpass Filter

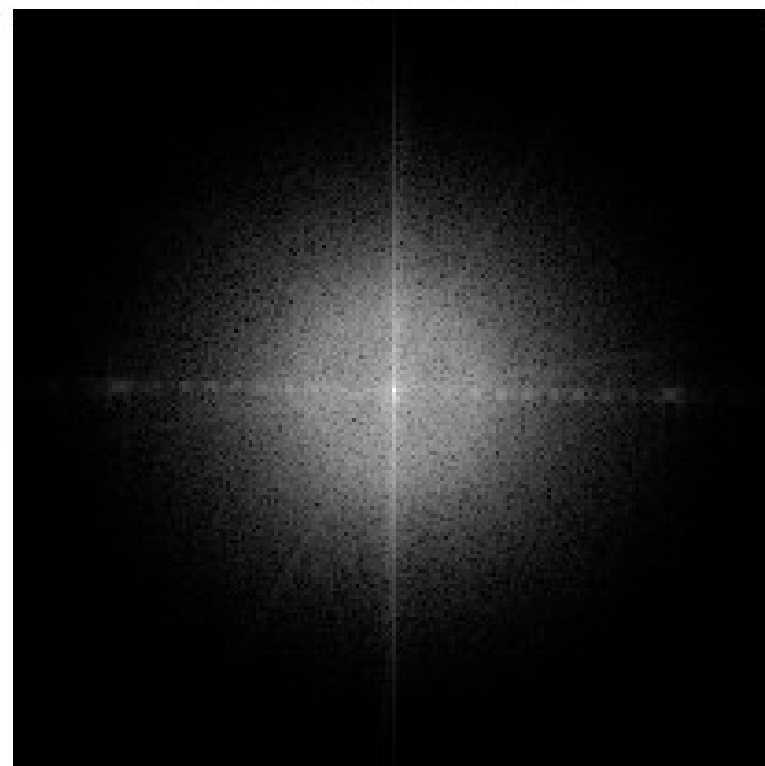
Original



FFT



Filtered FFT



Filtered Image





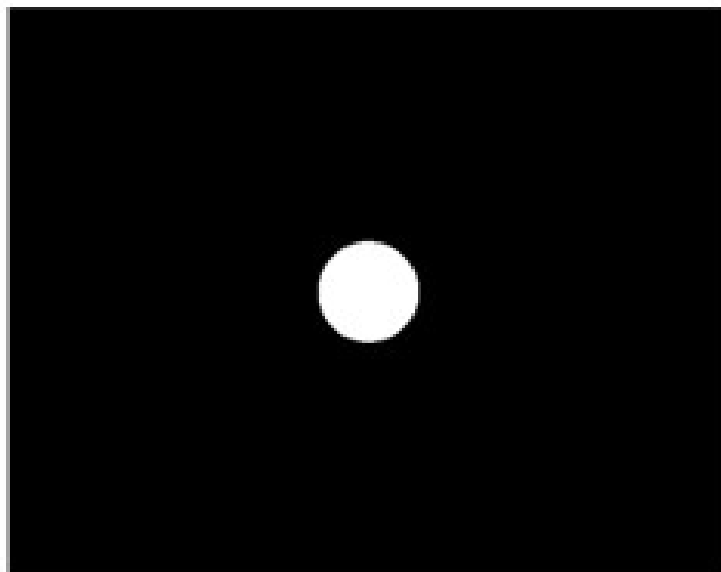
# Highpass Filters

- Highpass filter:

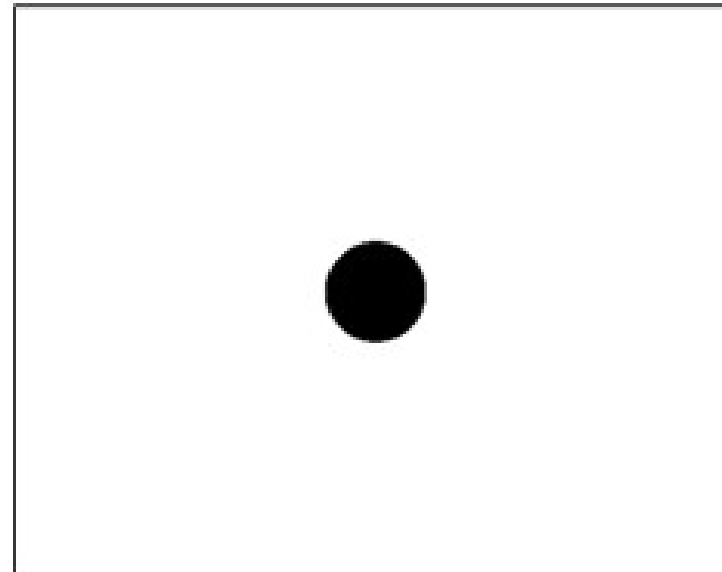
$$H_{HP} = 1 - H_{LP}$$

- High-frequency emphasis filter:

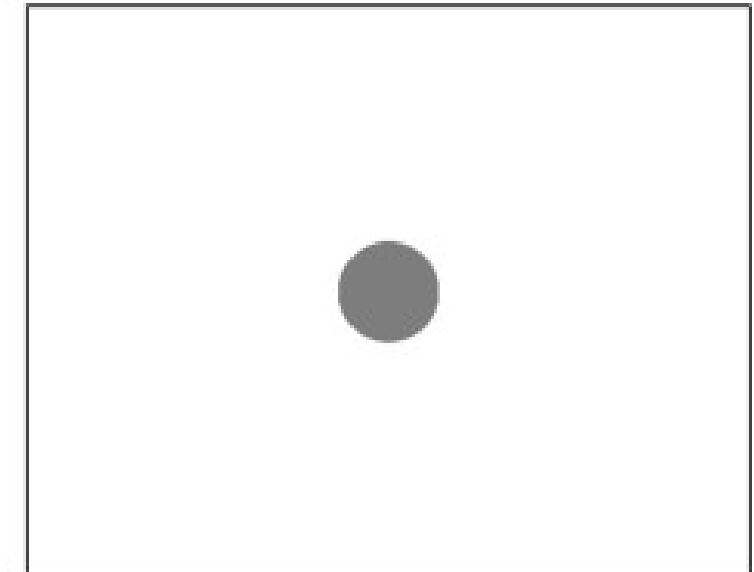
$$H_{HFE} = a + bH_{HP} \quad \text{with } a < b$$



$H_{LP,ideal}$



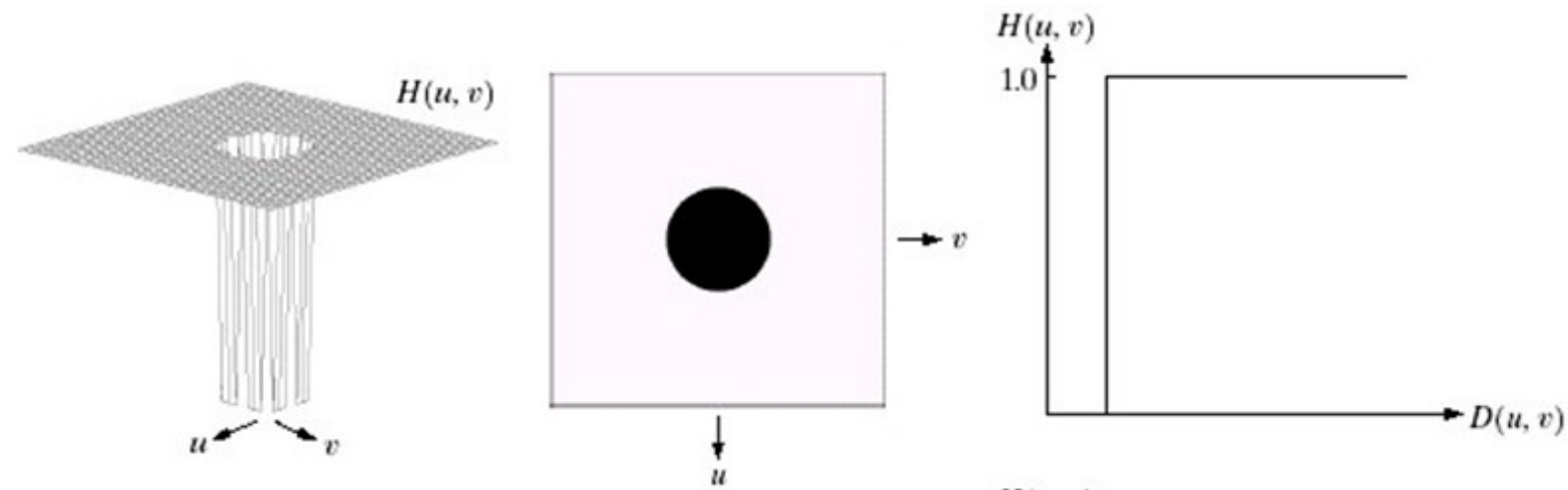
$H_{HP,ideal}$



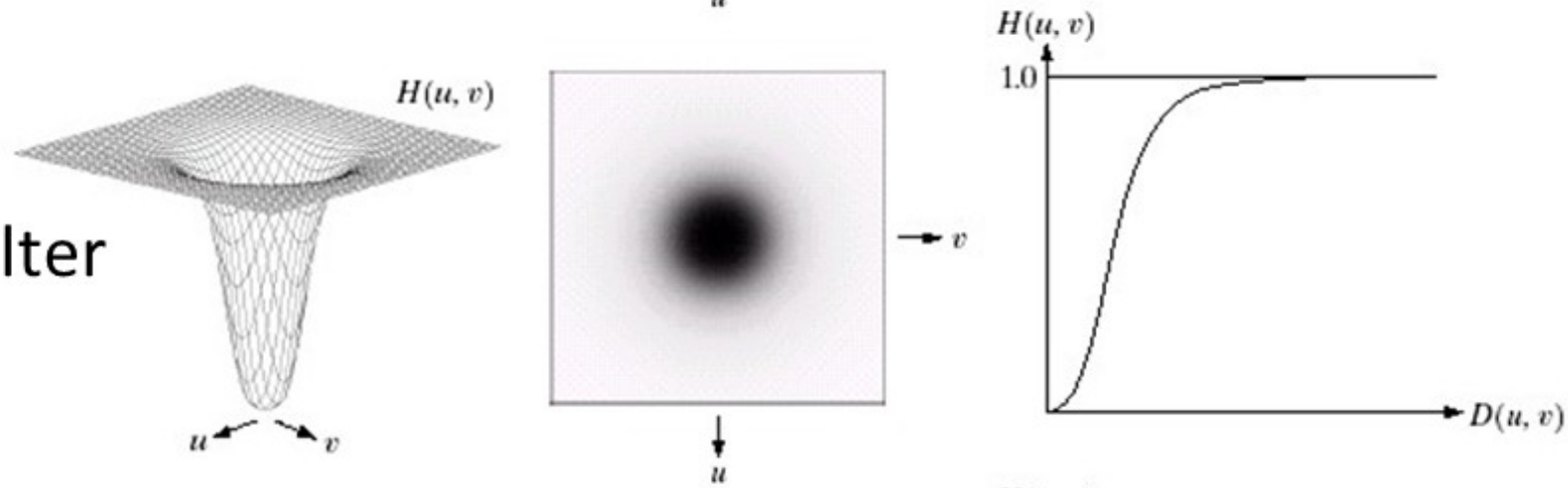
$H_{HFE,ideal}$

# Highpass Filters

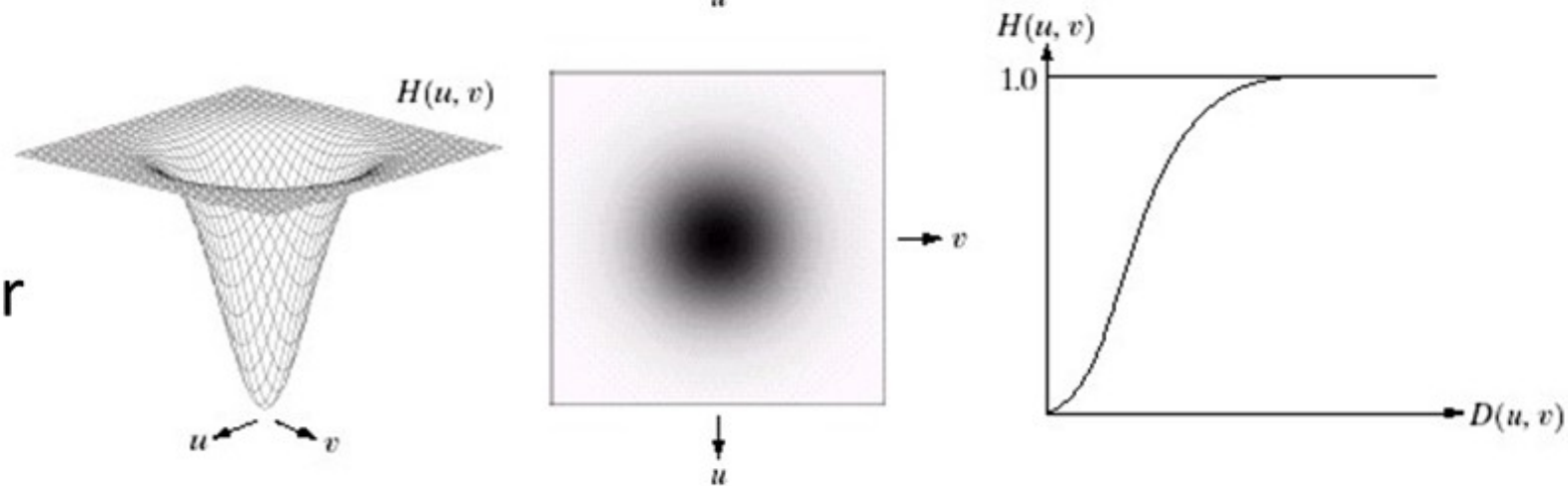
Ideal highpass filter



Butterworth highpass filter



Gaussian highpass filter



a b c  
d e f  
g h i

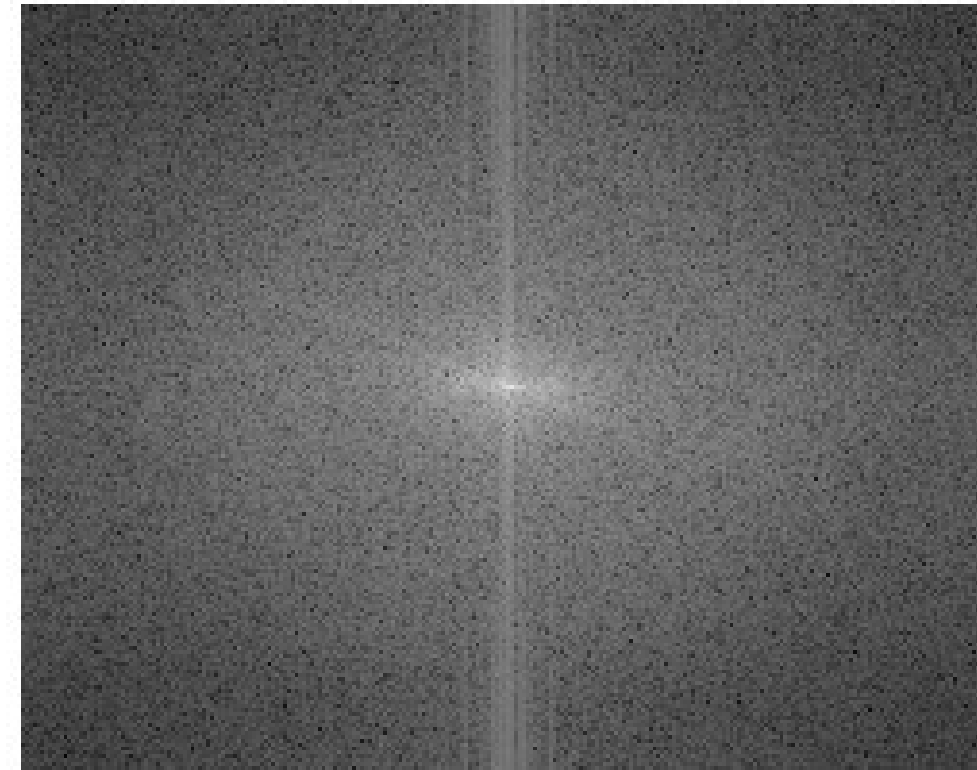
**FIGURE 4.22** Top row: Perspective plot, image representation, and cross section of a typical ideal highpass filter. Middle and bottom rows: The same sequence for typical Butterworth and Gaussian highpass filters.

# Exercise 4 – Butterworth Highpass Filter

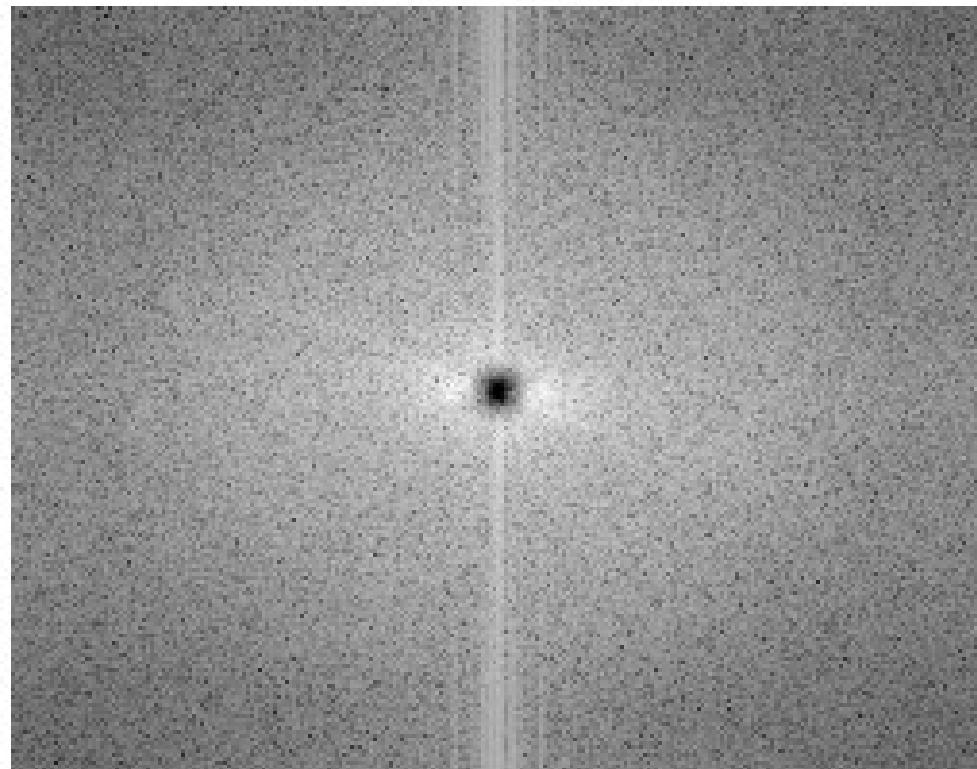
Original



FFT



Filtered FFT



Filtered Image

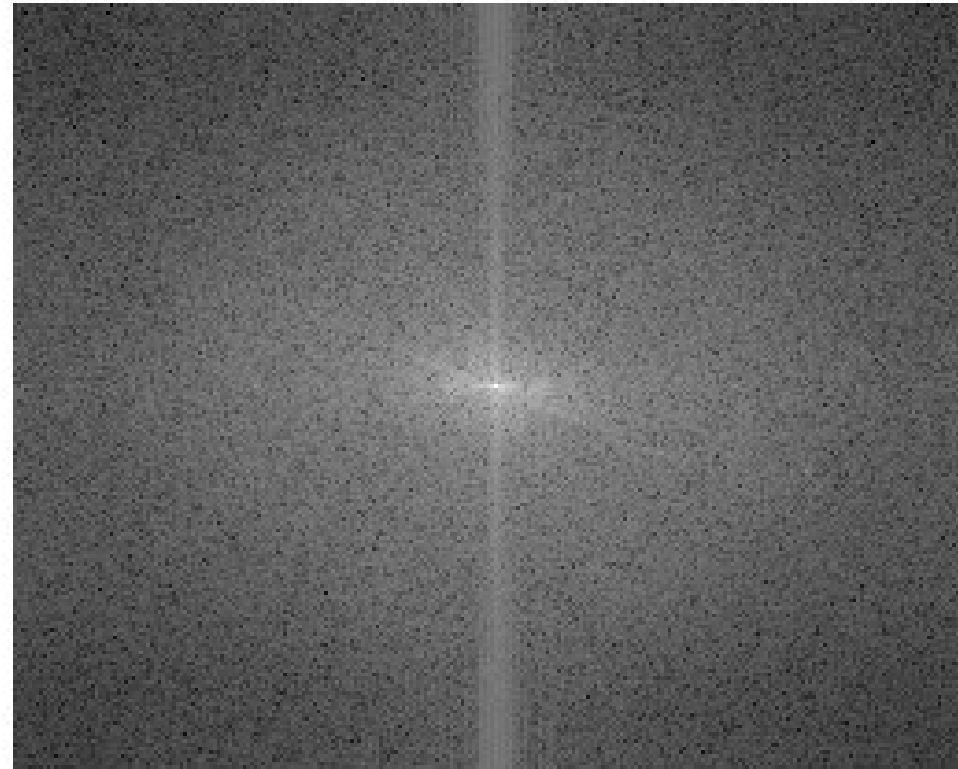


# Exercise 4 - Butterworth High Frequency Emphasis Filter

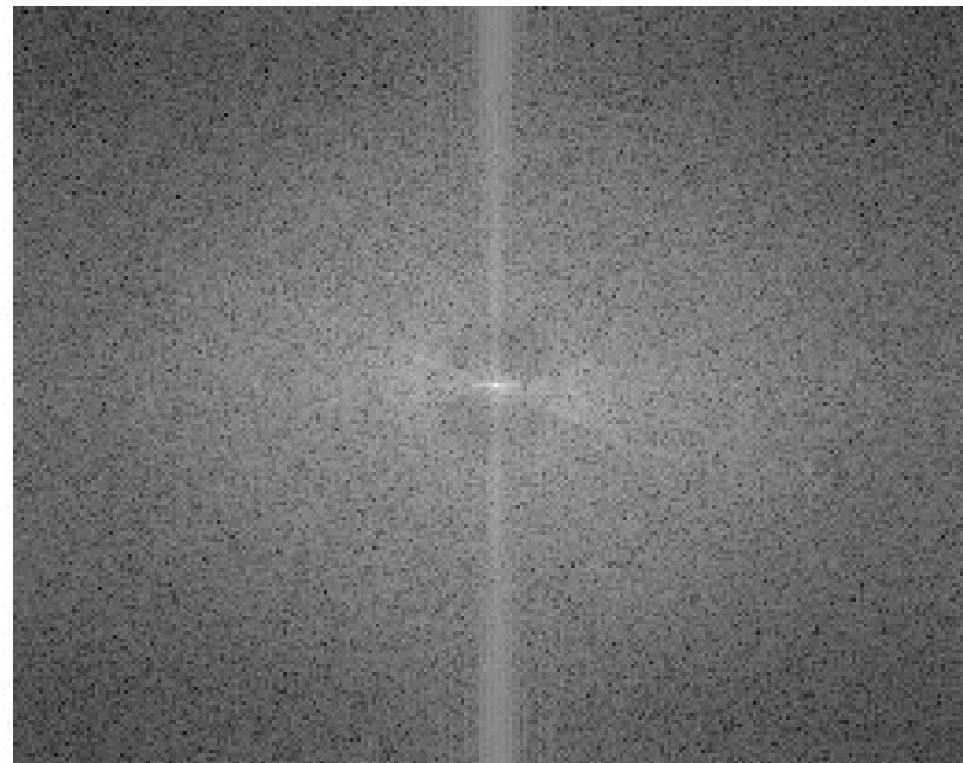
Original



FFT



Filtered FFT



Filtered Image

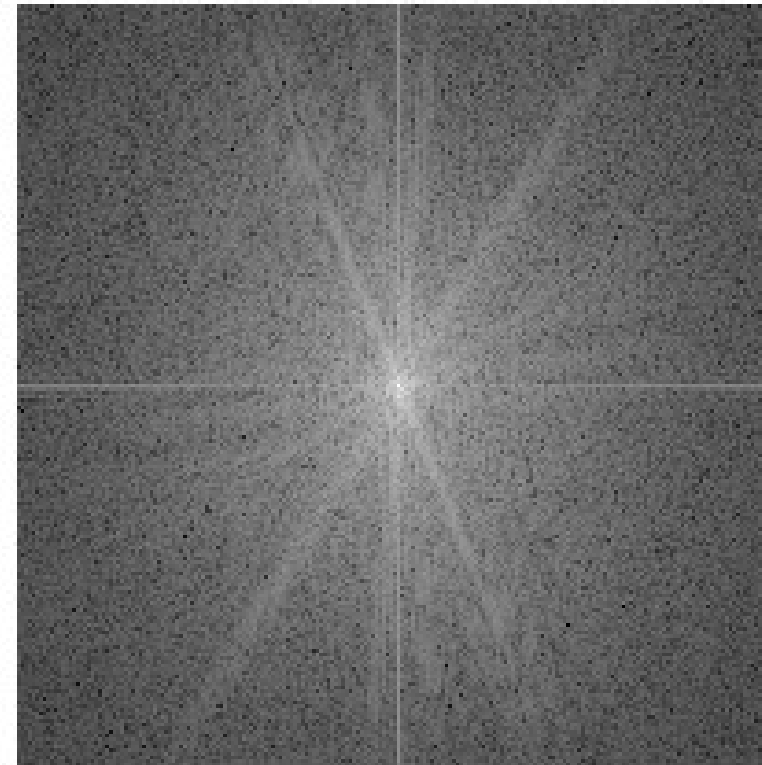


# Exercise 5 – Butterworth Notch Filter

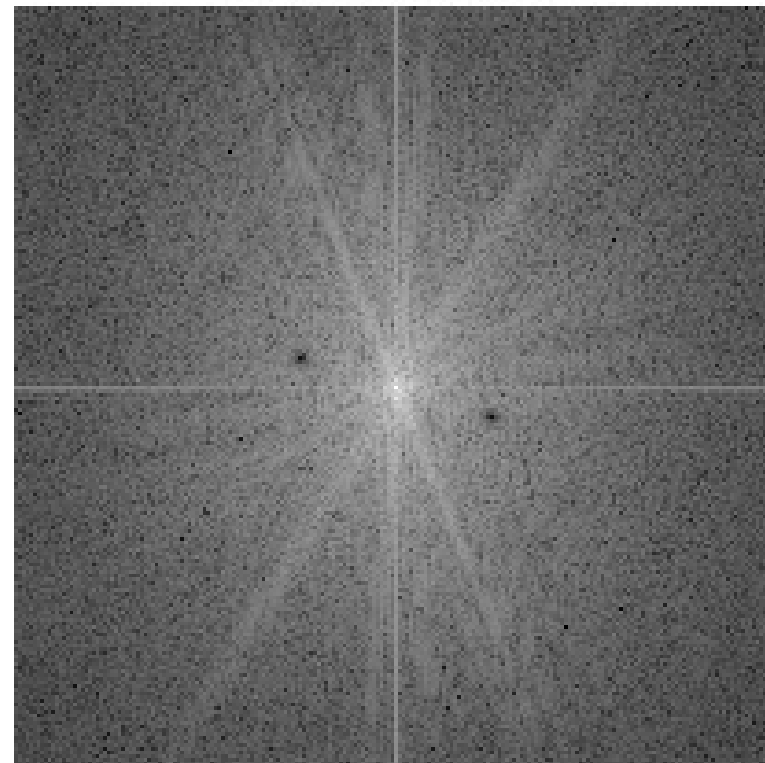
Original



FFT



Filtered FFT



Filtered Image



# TIPS & TRICKS

- **Deliver your working code before next session**
  - *Opdrachten* on Ufora
  - 1 python script named **labo#.py**
  - README.txt (see example)
  - **output.zip** with every output image

# TIPS & TRICKS

- **Questions or need help? Contact us at:**

Gianni.Allebosch@UGent.be AND Martin.Dimitrievski@UGent.be



# GOOD LUCK!