

22485 Medical Imaging systems

Lecture: Two-dimensional signal analysis

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Topic of today: Two-dimensional signal analysis

1. 2D Fourier transforms
 - (a) Relation to 1D Fourier transforms
 - (b) Examples and filtration
2. Hankel transform
3. Sampling of images
 - (a) Spatial sampling frequencies
 - (b) Discrete Fourier transforms
 - (c) Circular convolution
4. Discussion of exercise 4 about spectral velocity estimation
5. Questions for assignments

Reading material: L. Prince & J. M. Links: Medical imaging signals and systems, Pearson Prentice Hall Bioengineering, 2006 or 2015, Chapter 2 and 3.

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One dimensional Fourier transforms

Fourier transforms:

$$G(f) = \int_{-\infty}^{+\infty} g(t)e^{-j2\pi ft} dt$$

Inverse Fourier transforms:

$$g(t) = \int_{-\infty}^{+\infty} G(f)e^{j2\pi ft} df$$

If period is T then the frequency is $f_0 = 1/T$ in Hz or $1/s = s^{-1}$

Two dimensional Fourier transforms

Fourier transforms:

$$F(u, v) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y)e^{-j2\pi(ux+vy)} dx dy$$

Inverse Fourier transforms:

$$f(x, y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} F(u, v)e^{j2\pi(ux+vy)} du dv$$

If duration is X then the frequency is $f_x = 1/X$ in $1/m = m^{-1}$

Properties of two dimensional Fourier transforms

Complex: $F(u, v) = R(u, v) + jI(u, v)$

Magnitude and phase: $F(u, v) = M(u, v)e^{-j\phi(u, v)}$

$$M(u, v) = \sqrt{R(u, v)^2 + I(u, v)^2}$$

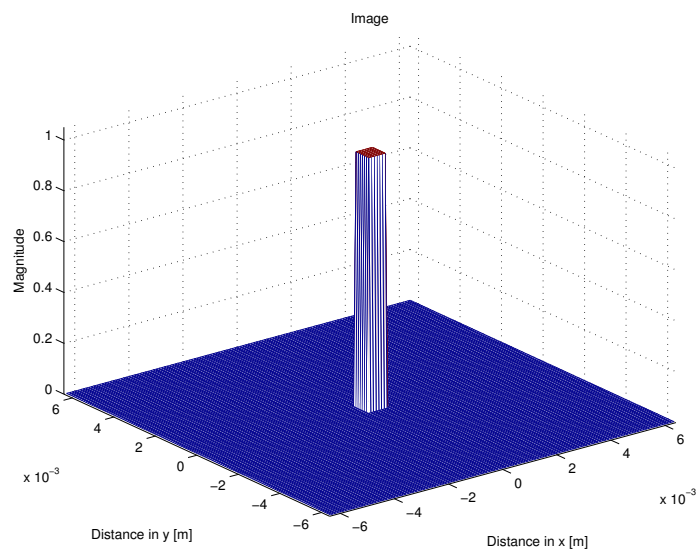
$$\phi(u, v) = \arctan(I(u, v), R(u, v))$$

Calculation of Fourier transforms:

$$F(u, v) = \int_{-\infty}^{+\infty} \left[\int_{-\infty}^{+\infty} f(x, y) e^{-j2\pi ux} dx \right] e^{-j2\pi vy} dy$$

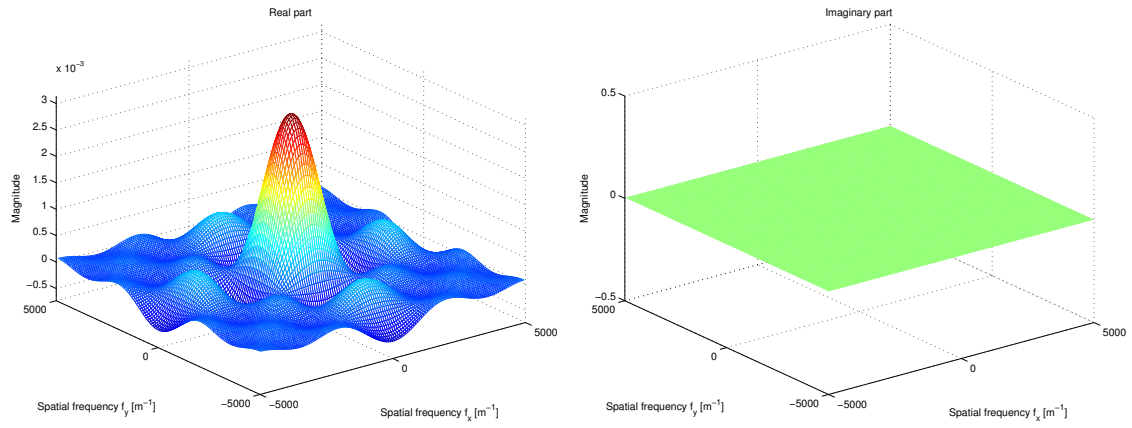
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Two dimensional square



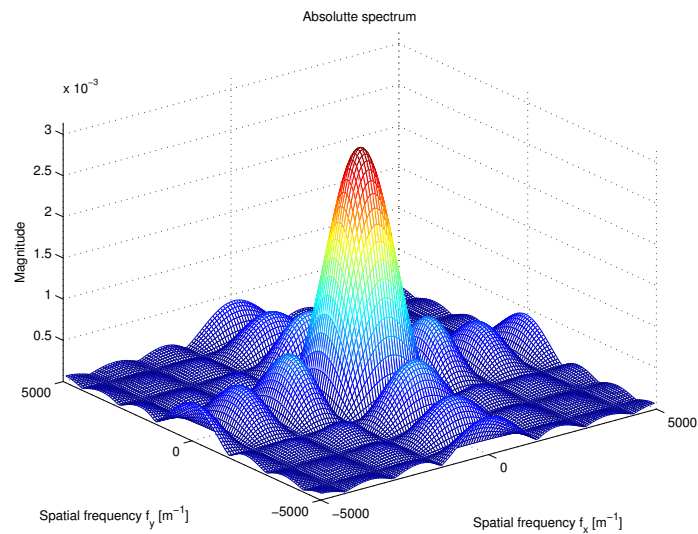
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Two dimensional Fourier transforms - real and imaginary part



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Two dimensional Fourier transforms - Magnitude



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Properties of two dimensional Fourier transforms

Linearity: $af_1(x, y) + bf_2(x, y) \leftrightarrow aF_1(u, v) + bF_2(u, v)$

Shift: $f(x - a, y - b) \leftrightarrow F(u, v)e^{-j2\pi(ua+vb)}$

Convolution: $f_1(x, y) * f_2(x, y) \leftrightarrow F_1(u, v)F_2(u, v)$

$$f_1(x, y) * f_2(x, y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_1(\alpha, \beta) f_2(x - \alpha, y - \beta) d\alpha d\beta$$

Multiplication: $f_1(x, y)f_2(x, y) \leftrightarrow F_1(u, v) * F_2(u, v)$

Separable image and Fourier transform

Image: $f(x, y) = f_x(x)f_y(y) \leftrightarrow F(u, v) = F_x(u)F_y(v)$

Example for square:

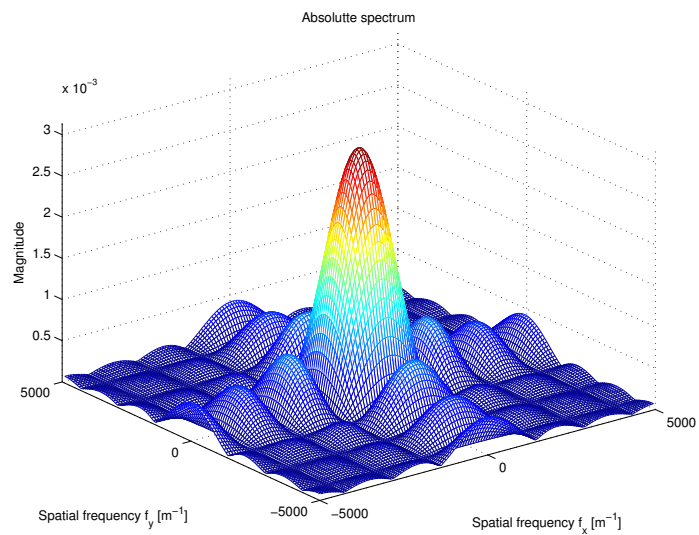
$$f_x(x) = 1, |x| < k_x, \text{ else } 0 \leftrightarrow F_x(u) = 2k_x \frac{\sin \pi u 2k_x}{\pi u 2k_x}$$

$$f_y(y) = 1, |y| < k_y, \text{ else } 0 \leftrightarrow F_y(v) = 2k_y \frac{\sin \pi v 2k_y}{\pi v 2k_y}$$

Spectrum for image:

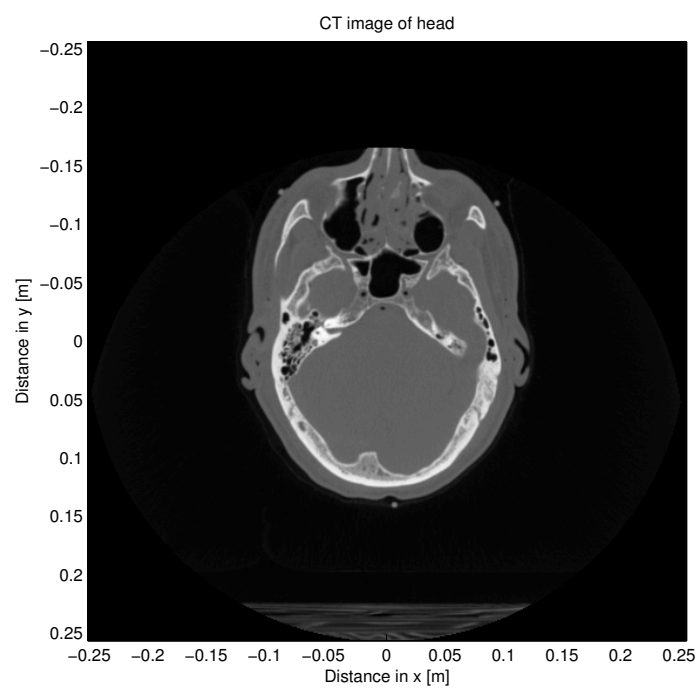
$$f(x, y) = f_x(x)f_y(y) \leftrightarrow F(u, v) = F_x(u)F_y(v) = 2k_x \frac{\sin \pi u 2k_x}{\pi u 2k_x} 2k_y \frac{\sin \pi v 2k_y}{\pi v 2k_y}$$

Two dimensional Fourier transforms - Magnitude



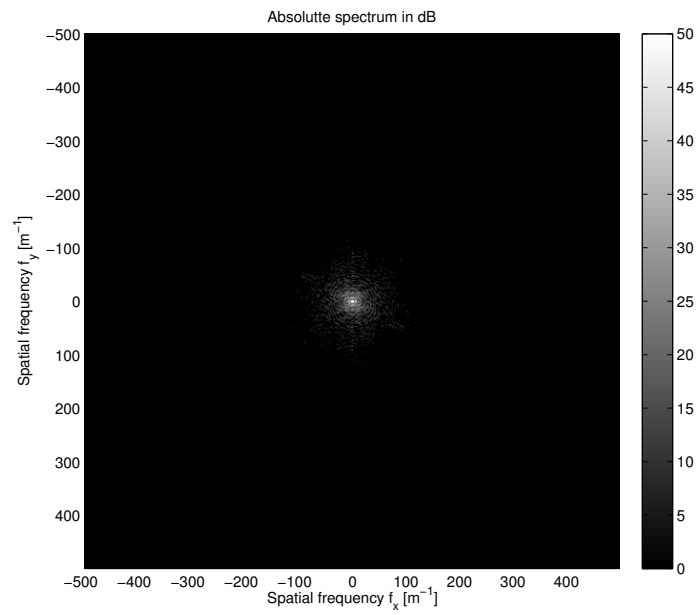
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CT head image



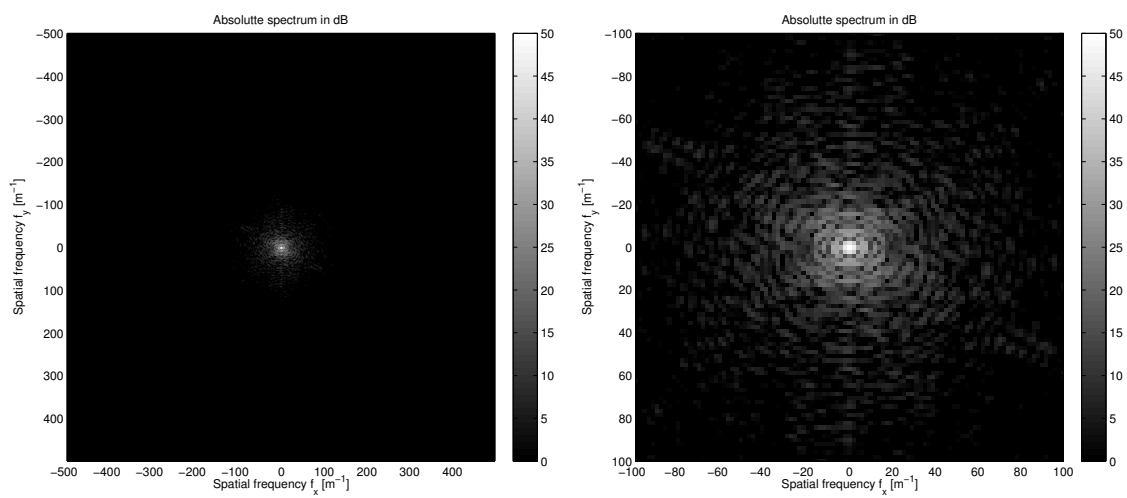
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Two dimensional Fourier transforms - Magnitude



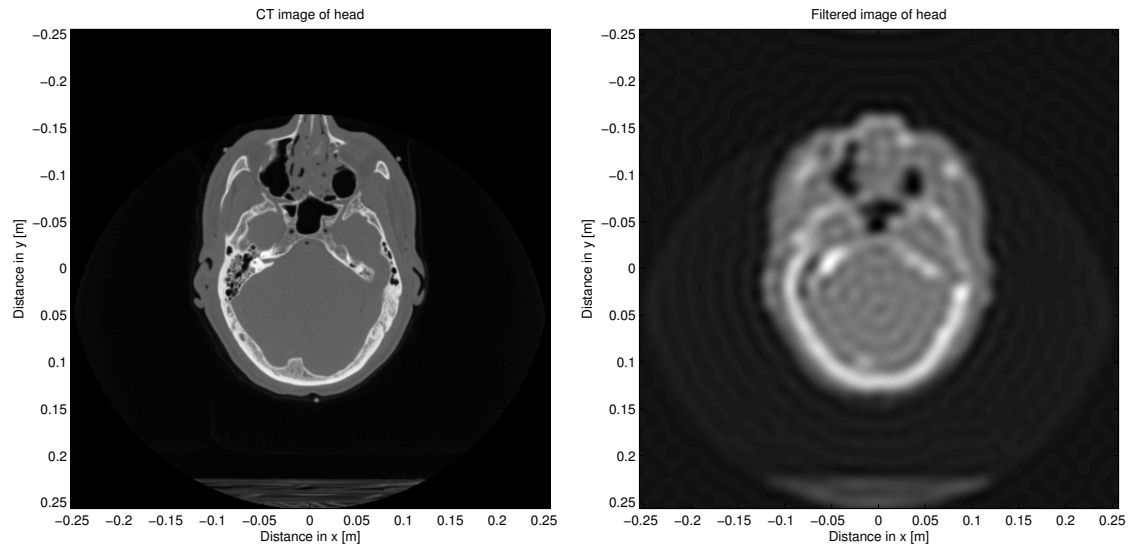
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Two dimensional Fourier transforms - Magnitude zoomed



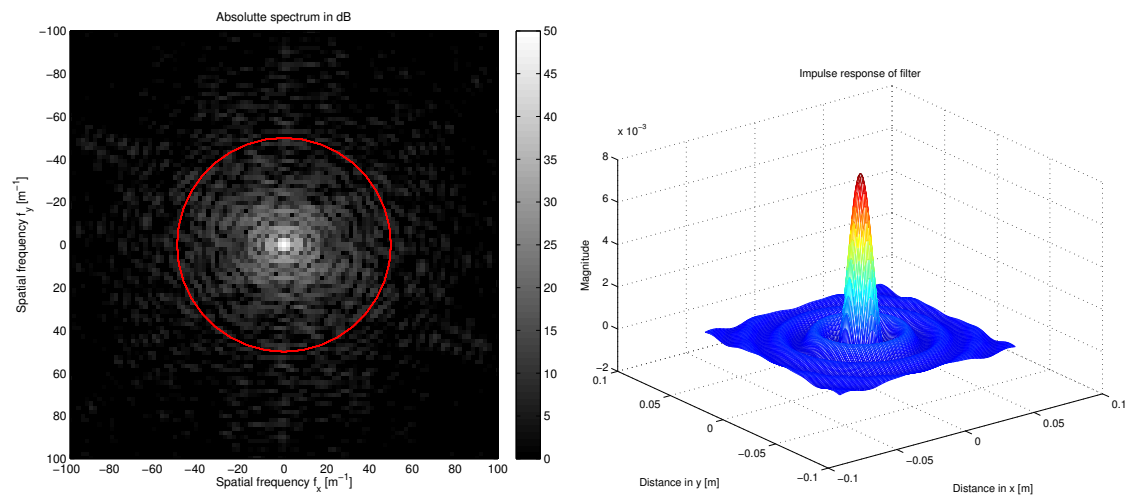
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Low-pass filtration of head image



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Low-pass filter for head image

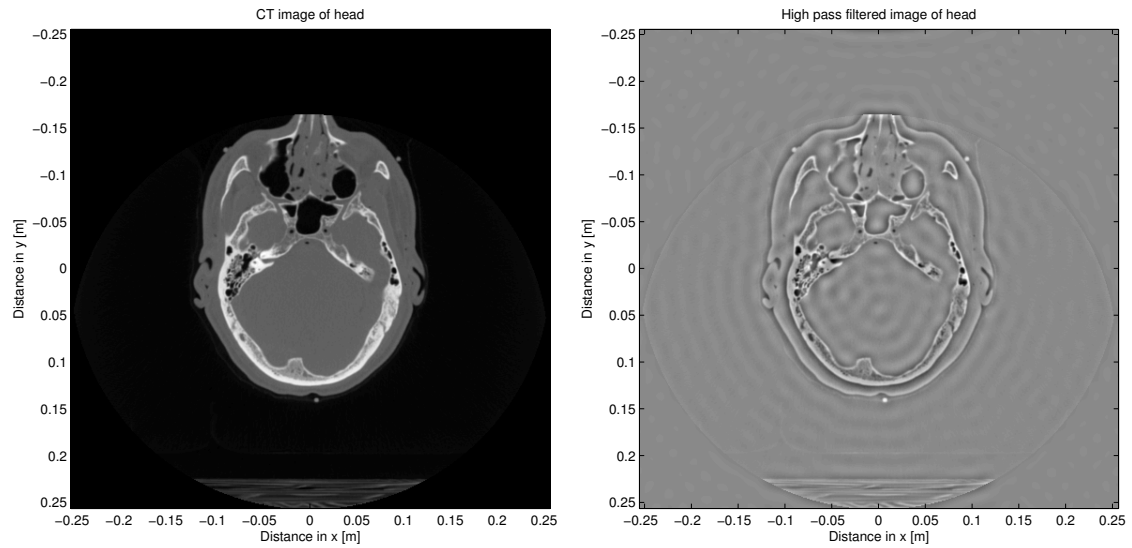


Spectrum in frequency domain

Impulse response

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High-pass filtration of head image



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Circular symmetric images - Hankel transform

Circular symmetric image: $f(x, y) = f(r, \phi) = f(r)$

Fourier transform circular symmetric: $F(u, v) = F(q, \Theta) = F(q)$

Given by the Hankel transform:

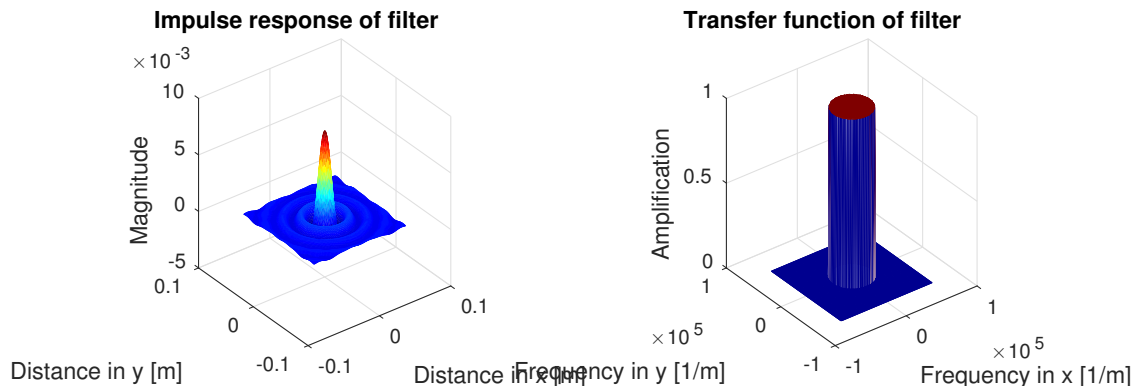
$$F(q) = 2\pi \int_0^{+\infty} r f(r) J_0(2\pi qr) dr$$

Bessel function of first kind:

$$J_0(r) = \frac{1}{\pi} \int_0^{\pi} \cos(r \sin \phi) d\phi$$

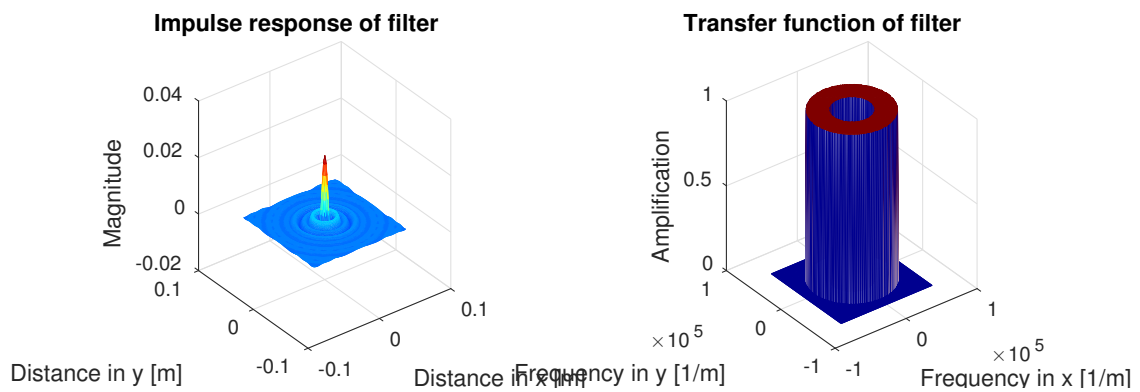
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Circular symmetry - Hankel transform



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Circular symmetry - Hankel transform



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List of Hankel transforms

$$\exp(-\pi r^2) \leftrightarrow \exp(-\pi q^2)$$

$$1 \leftrightarrow \frac{\delta(q)}{\pi q}$$

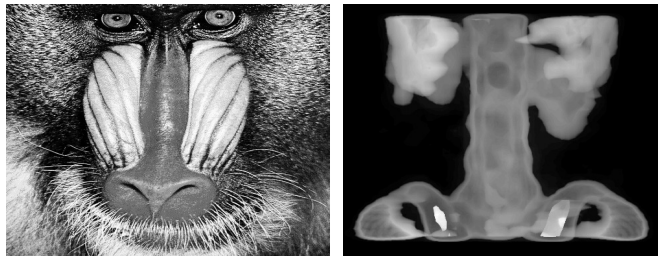
$$\delta(r - a) \leftrightarrow 2\pi J_0(2\pi qa)$$

$$\text{sinc}(r) \leftrightarrow \frac{2\text{rect}(q)}{\pi\sqrt{1 - 4q^2}}$$

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Discussion after break

Consider what image you will see, if you combine the phase from one image with the amplitude from another image.

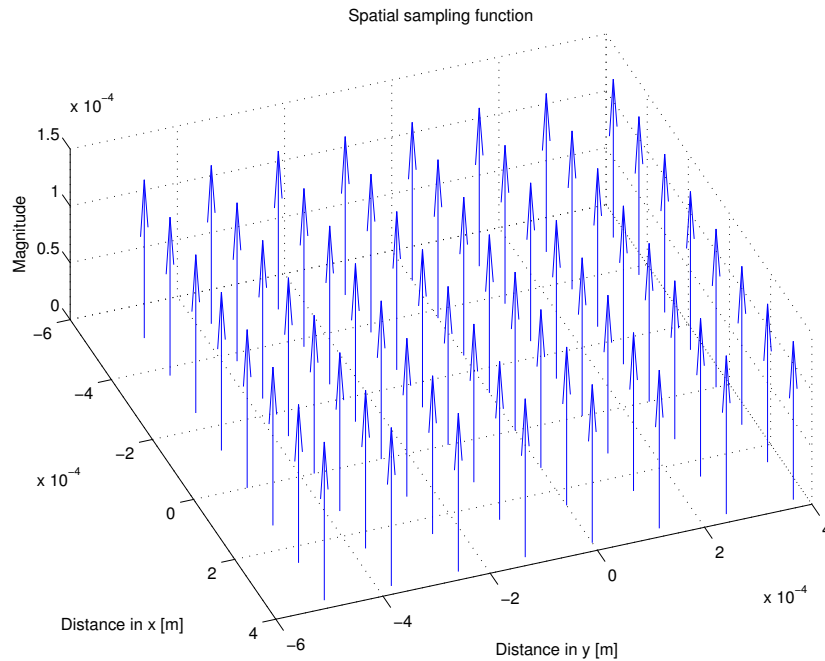


1. What is likely the amplitude spectrum of both images?
2. How would the phase be?
3. What is the combination?

for_10_2d_signals/matlab_demo/phase_demo.m

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Sampling function



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Sampling

2D image of δ -functions:

$$s(x, y) = \sum_{j=-\infty}^{+\infty} \sum_{k=-\infty}^{+\infty} \delta(x - j\Delta x, y - k\Delta y)$$

Spatial sampling frequencies: $f_{sx} = 1/\Delta x$, $f_{sy} = 1/\Delta y$

Sampled image:

$$f_s(x, y) = f(x, y)s(x, y) \leftrightarrow F(u, v) * S(u, v)$$

Fourier transform of sampling function:

$$s(x, y) \leftrightarrow S(u, v) = \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{+\infty} \sum_{n=-\infty}^{+\infty} \delta(u - m f_{sx}, v - n f_{sy})$$

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Spectrum of sampled image

Remember

$$F(u, v) * \delta(u - mf_{sx}, v - nf_{sy}) = F(u - mf_{sx}, v - nf_{sy})$$

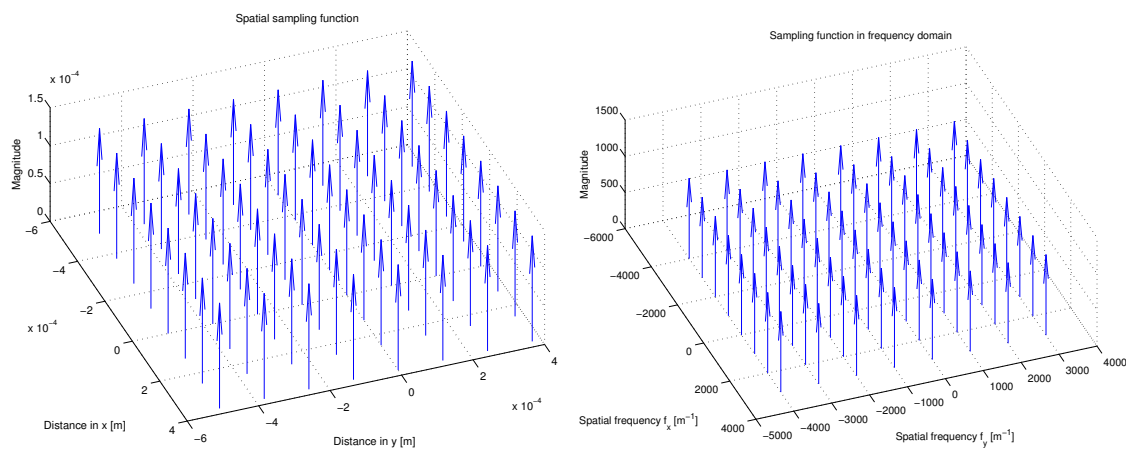
Spectrum

$$\begin{aligned} f_s(x, y) = f(x, y)s(x, y) &\leftrightarrow F(u, v) * S(u, v) \\ &= F(u, v) * \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{+\infty} \sum_{n=-\infty}^{+\infty} \delta(u - mf_{sx}, v - nf_{sy}) \\ F_s(u, v) &= \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{+\infty} \sum_{n=-\infty}^{+\infty} F(u - mf_{sx}, v - nf_{sy}) \end{aligned}$$

Spectrum repeats itself with a period of f_{sx} and f_{sy}

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Sampling function

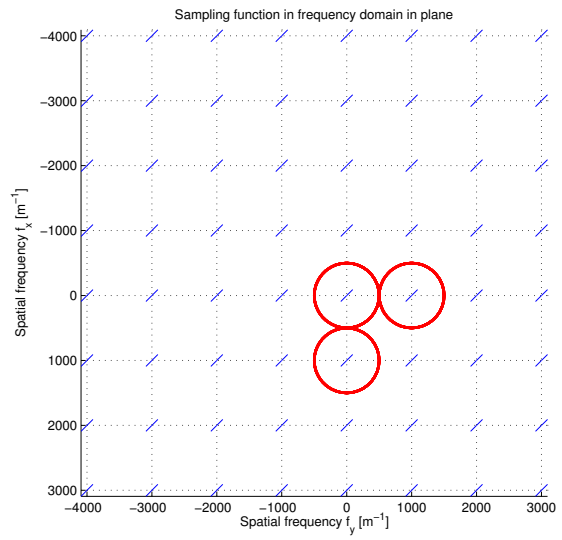
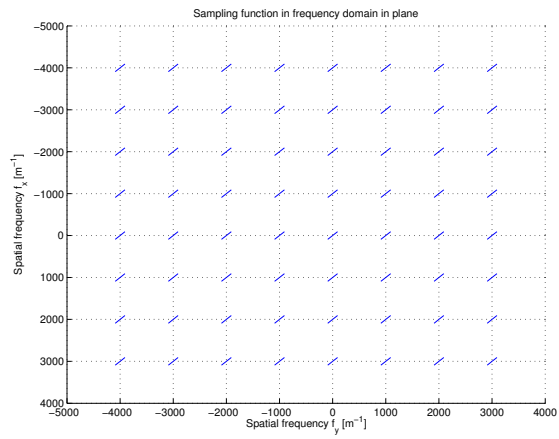


Spatial domain

Spectrum in frequency domain

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Sampling function



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Spatial Aliasing



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Discrete two dimensional Fourier transforms

Discrete Fourier transforms:

$$F(u_d, v_d) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m, n) e^{-j2\pi \left(\frac{mu_d}{M} + \frac{nv_d}{N} \right)}$$

Frequency variables are discrete: $u_d = 0..M - 1$ and $v_d = 0..N - 1$

Spectrum is discrete and periodic;

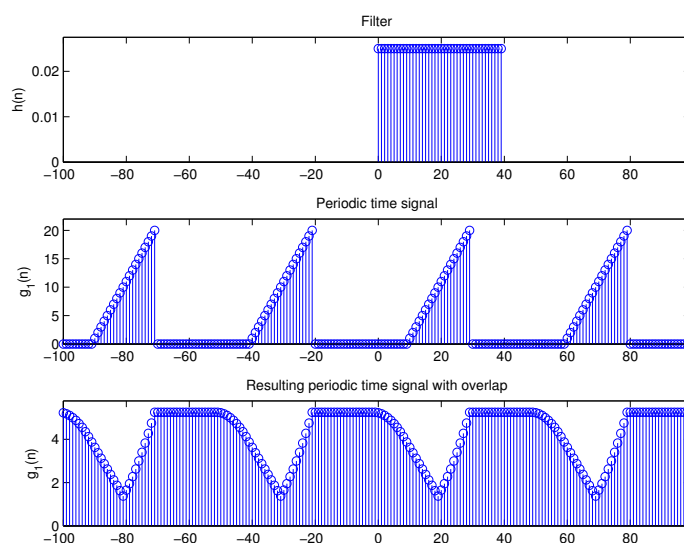
$$F(u_d, v_d) = F(M - u_d, N - v_d)$$

Inverse discrete Fourier transforms:

$$f(m, n) = \sum_{u_d=0}^{M-1} \sum_{v_d=0}^{N-1} F(u_d, v_d) e^{j2\pi \left(\frac{mu_d}{M} + \frac{nv_d}{N} \right)}$$

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Circular convolution

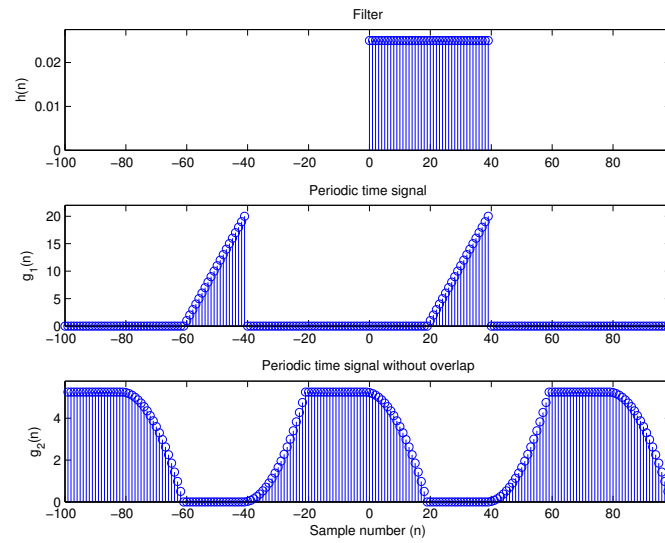


N_1 - Length signal 1, N_2 - Length signal 2, N - Transform length

Here: $N < N_1 + N_2 - 1$

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Circular convolution

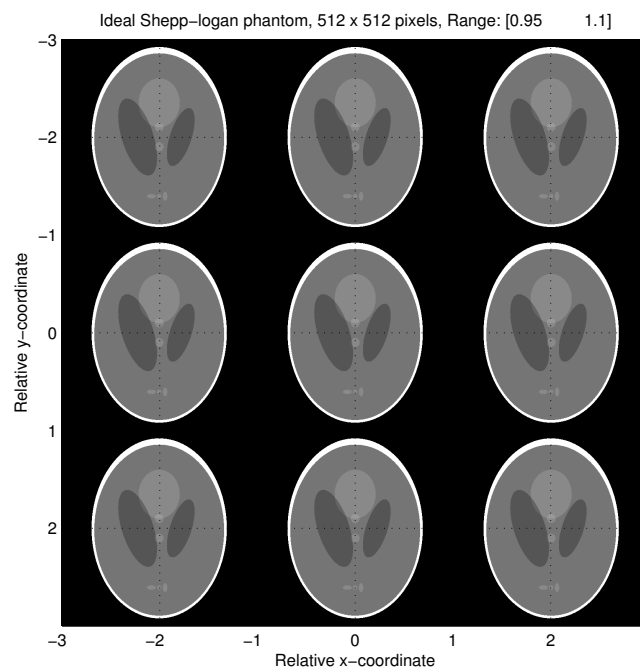


N_1 - Length signal 1, N_2 - Length signal 2, N - Transform length

Here: $N > N_1 + N_2 - 1$

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Circular convolution for Shepp-Logan phantom



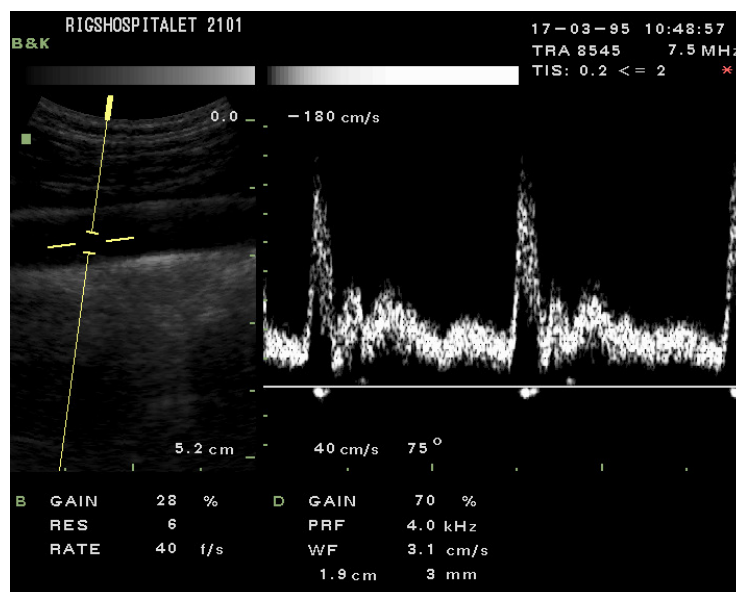
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Exercise 4: signal processing in pulsed wave system

1. Process receive signal to get complex data (load from file)
2. Divide into overlapping segments
3. Calculate power spectrum (apply compression)
4. Display the spectra as a function of time
5. Compare the spectra for different vessels

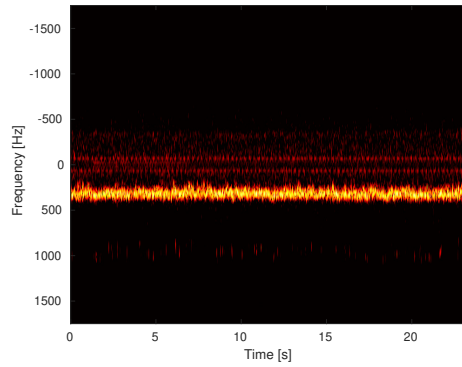
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Spectrogram from carotid artery



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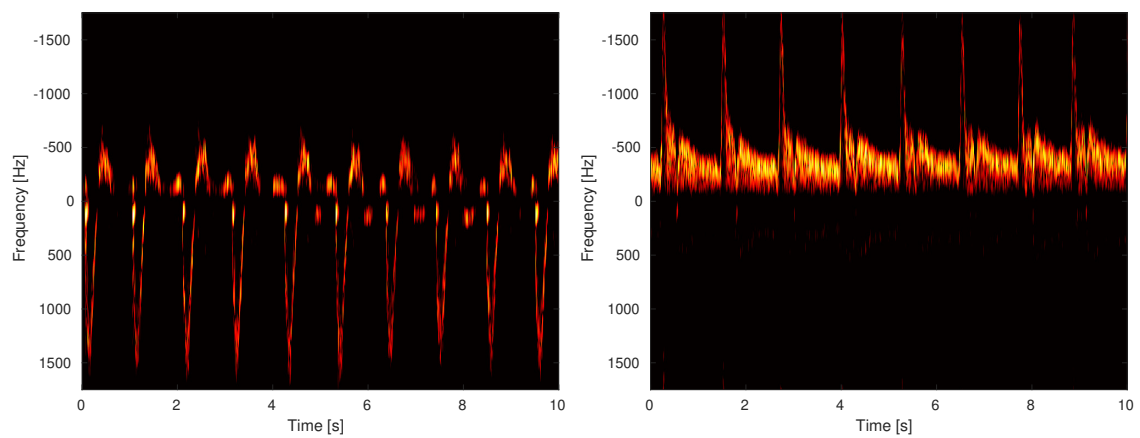
Spectrogram for phantom



- Spectrogram for phantom data with constant parabolic flow
- 30 dB dynamic range
- 128 samples segments every 1 ms to get smooth appearance
- Hanning windowing
- FFT calculated for 1024 samples to increase smoothness

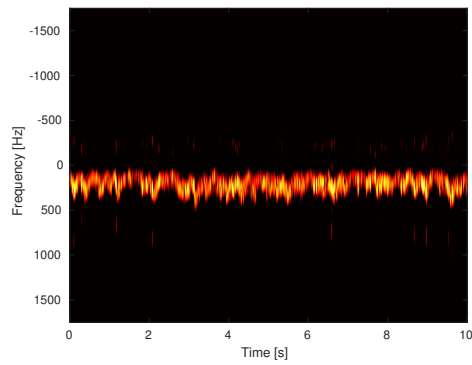
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Spectrograms for aorta (left) and carotis (right)



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Spectrogram for porta vein



- 30 dB dynamic range
- 128 samples segments every 1 ms to get smooth appearance
- Hanning windowing
- FFT calculated for 1024 samples to increase smoothness