

EDISP (2D sig)
(English) Digital Signal Processing
Two-dimensional signals & filters
lecture

May 31, 2017

Multidimensional signals

- ▶ Analogue K-D signal $x_a(t_1, t_2, t_3, \dots, t_K)$, t_k not necessarily time.
- ▶ Discretization (sampling) $\longrightarrow x(n_1, n_2, n_3, \dots, n_K)$, *some signals are already discrete!*
 \longrightarrow multidimensional discrete signal = multidimensional number series
- ▶ sampling periods $T_{s\ k} \longrightarrow$ sampling frequencies $f_{s\ k} = \frac{1}{T_{s\ k}}$ not necessarily equal; (e.g some scanners have different h & v resolutions)
if t_k is spatial, $f_{s\ k}$ is *spatial frequency*
- ▶ Examples
 - ▶ 2-D picture
 - ▶ linear antenna array (t_1 – discrete or continuous space, t_2 – continuous time)
 - ▶ pulsed radar signal (“slow time” and “fast time”)

ULA - Uniform Linear Array (of antennas)



SDMA (a.k.a. BDMA) - Space (Beam) Division Multiple Access

Smart LTE antenna



Two dimensions in classical radar

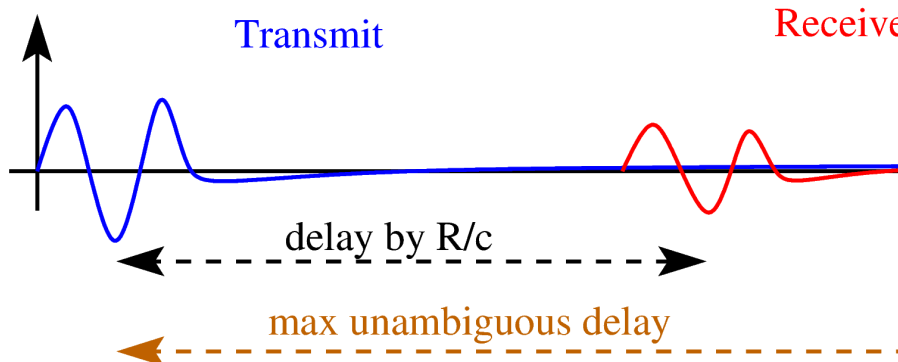


Image – a 2-D signal – and its 2-D DFT

- ▶ T_{s1} , T_{s2} – pixel dimensions; f_{s1} , f_{s2} – resolution (dpi, lines/mm)
- ▶ Fourier spectrum:

$$X(e^{j\theta_1}, e^{j\theta_2}) = \sum_{n_1=-\infty}^{+\infty} \sum_{n_2=-\infty}^{+\infty} x(n_1, n_2) e^{-jn_1\theta_1} e^{-jn_2\theta_2}$$

$\theta_k = \omega_k \cdot T_{sk} = \frac{2\pi f_k}{f_{sk}}$ – normalized angular frequency

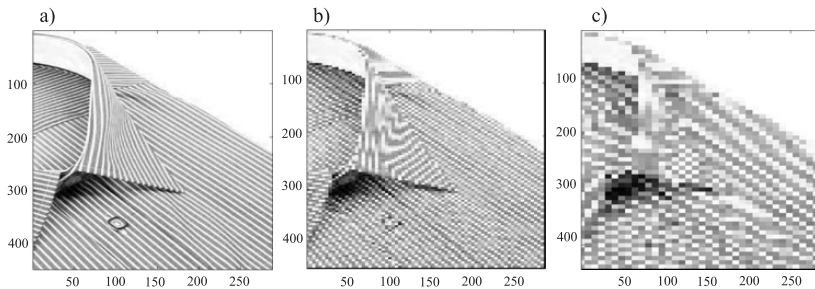
- ▶ finite picture \longrightarrow represented by a discrete spectrum

$$X(k_1, k_2) = \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} x(n_1, n_2) e^{-j2\pi k_1 n_1 / N_1} e^{-j2\pi k_2 n_2 / N_2}$$

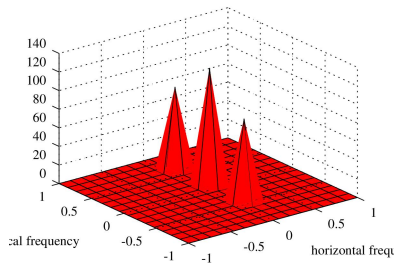
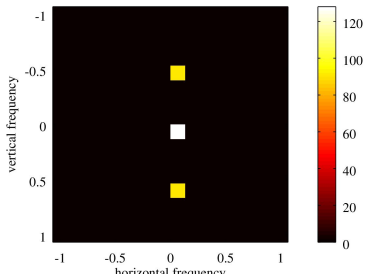
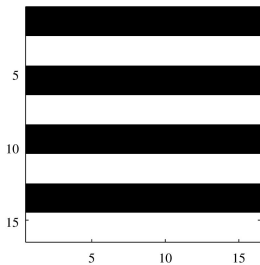
- ▶ reconstruction by a discrete Fourier series

$$x(n_1, n_2) = \frac{1}{N_1 N_2} \sum_{k_1=0}^{N_1-1} \sum_{k_2=0}^{N_2-1} X(k_1, k_2) e^{j2\pi k_1 n_1 / N_1} e^{j2\pi k_2 n_2 / N_2}$$

image sampling example



2-D DFT example



2-D LTI systems

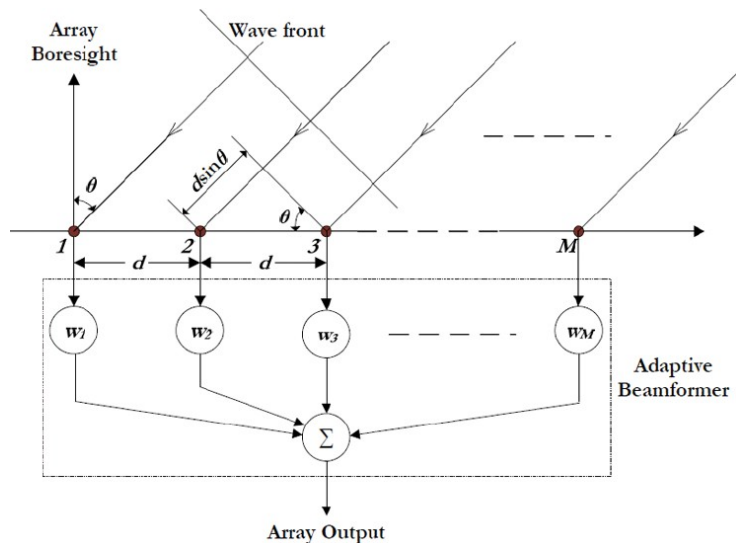
Linear and (*Time = shift in n_k*) Invariant (we extend the definition for 1-D systems)

- ▶ allows for analysis by impulse response
(unit impulse: $\delta(m, n) = 1$ if $m = n = 0$, $= 1$ otherwise)
impulse response is sometimes called Point Spread Function – PSF (especially in optics)
- ▶ causality – meaningful if one of dimensions is time-related
- ▶ delay - in spatial dimension, zero-delay filter is the best
- ▶ 2-D convolution (linear filtering)

$$y(n_1, n_2) = \sum_{m_1=-\infty}^{+\infty} \sum_{m_2=-\infty}^{+\infty} h(m_1, m_2) \cdot x(n_1 - m_1, n_2 - m_2)$$

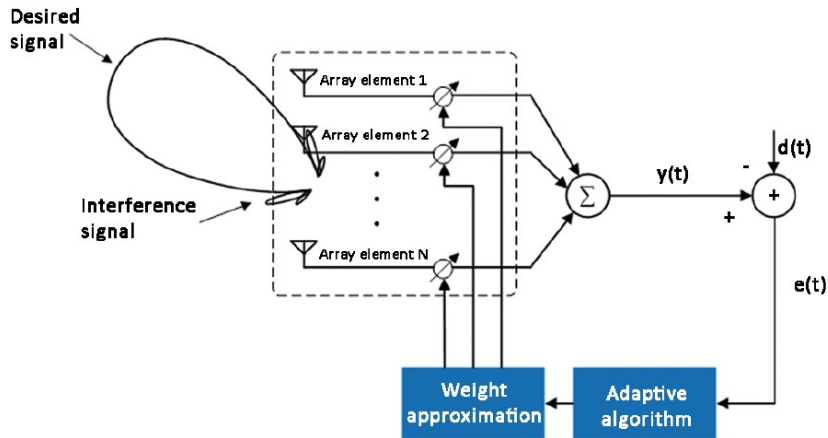
- ▶ if $h(m_1, m_2) = h_1(m_1) \cdot h_2(m_2)$, we may decompose 2-D filtering into 2x(1-D) (important for long impulse responses)

Beamforming

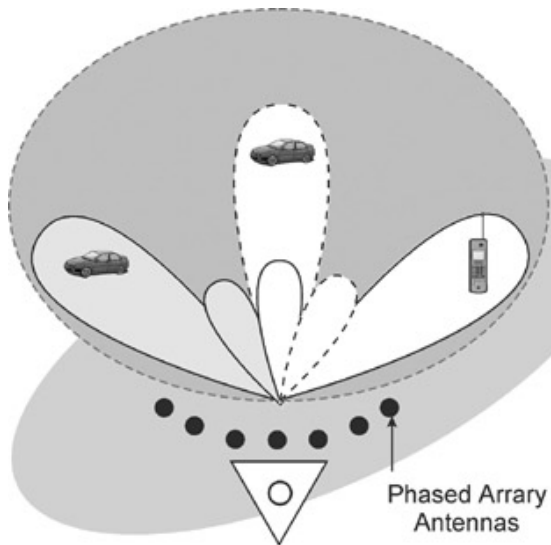


$$x_k(t) = A(t) \cos(\Omega t + \frac{nd}{\lambda} \sin \theta) \longrightarrow x_k^{bb}(t) = A(t) e^{j \frac{nd}{\lambda} \sin \theta}$$

Beamforming

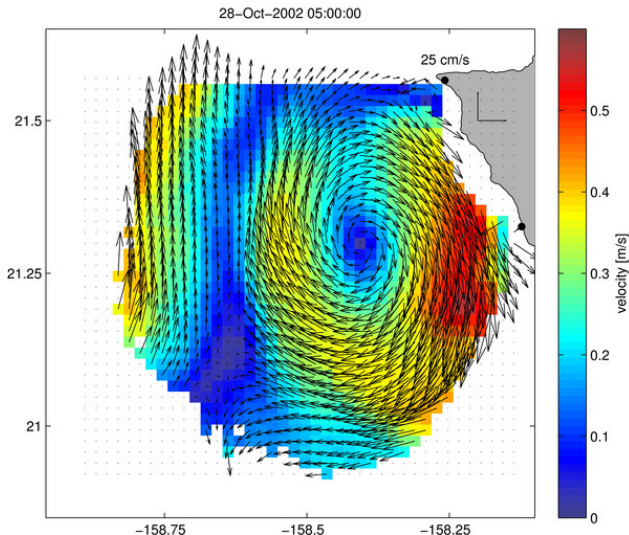


Application - beamforming in SDMA



Application - ocean currents monitoring (WERA)

WERA is a HF radar using electromagnetic waves between 6 and 30 MHz (50 m to 10 m wave length) to measure surface current velocities, ocean wave height (spectra) and wind.



Images - practical remarks

We concentrate on monochrome images (B/W photos, print, raw images in medical, radar, sonar, satellite technology). Color images add some complexity, unimportant for the signal processing basics.

Understanding frequency concept:

- ▶ Horizontal frequency - changes in the horizontal direction (trunks in the forest)
- ▶ Vertical frequency - changes in the vertical direction (horizon)
- ▶ High frequencies - sharp edges, small features
- ▶ Low frequencies - soft edges, large areas of same intensity

Filtering

Why:

- ▶ resizing by interpolation/decimation (LP)
- ▶ removing noise or interference (LP)
- ▶ sharpening images (slightly HP)
- ▶ detecting edges (HP)

Filtering methods:

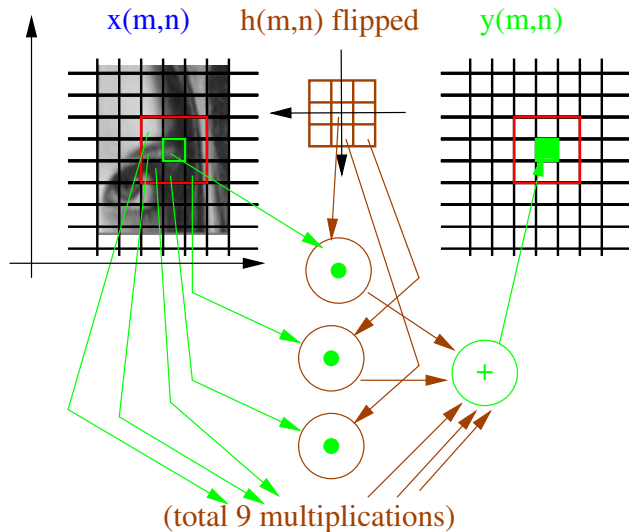
- ▶ Spatial (image) domain: convolution=weighted average of neighboring pixels
- ▶ Frequency domain: masking out (zeroing) parts of the 2D spectrum
- ▶ Nonlinear filters: some nonlinear manipulation on the set of neighboring pixels (e.g median)

How to behave at the image boundary with spatial filters?

- ▶ assume zeros outside (effect:dark areas at the boundaries)
- ▶ repeat last data row/column
- ▶ circular symmetry (effect: sky under the ground?)
- ▶ mirror symmetry (best for photo type images)

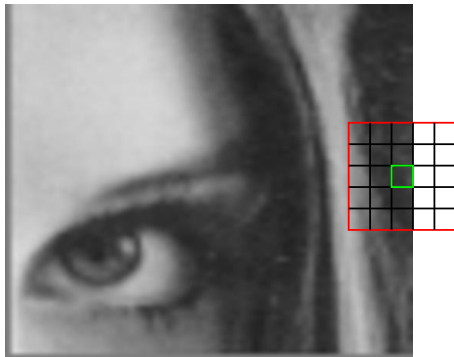
Spatial filter (2D convolution)

Operation on some vicinity of the current pixel.



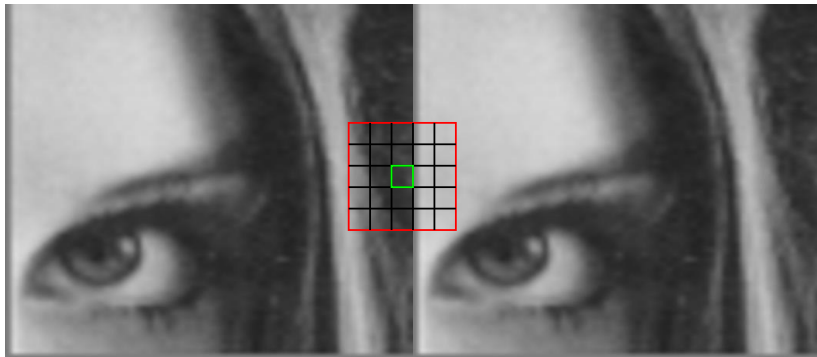
$$y(m,n) = \sum_{i=-\infty}^{+\infty} \sum_{j=-\infty}^{+\infty} x(i,j) \cdot h(m-i, n-j)$$

Edge problem



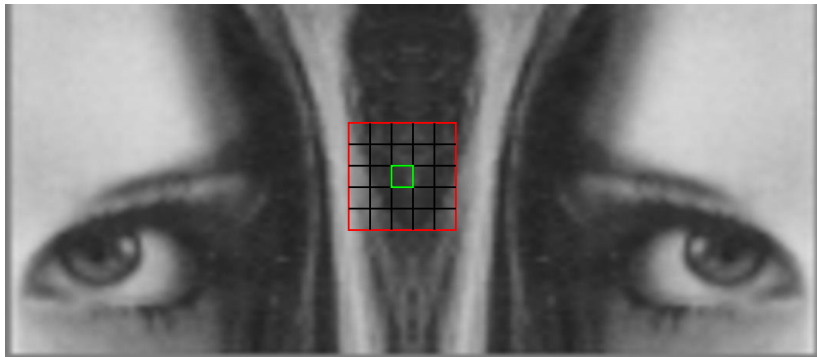
Assume zero outside

Edge problem



Assume circular copy

Edge problem



Assume flipped (mirror) copy

Spectral domain filtering (by FFT)

Operation on whole picture.

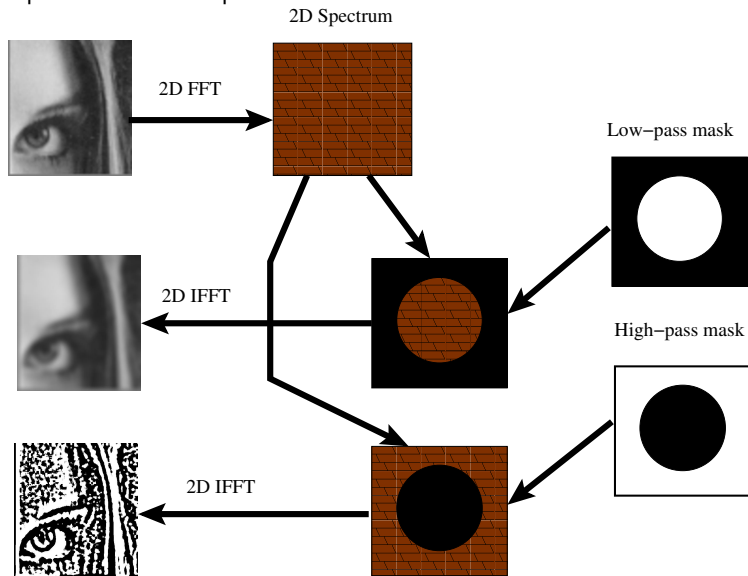


Image denoising

- ▶ AWGN - linear filtering
- ▶ impulsive noise (“salt and pepper”) - nonlinear methods

Median filter (nonlinear)

- ▶ removing impulses
- ▶ preserving edges
- ▶ method: replace a pixel with median from its vicinity (3x3, 5x5, ...)

Median filter

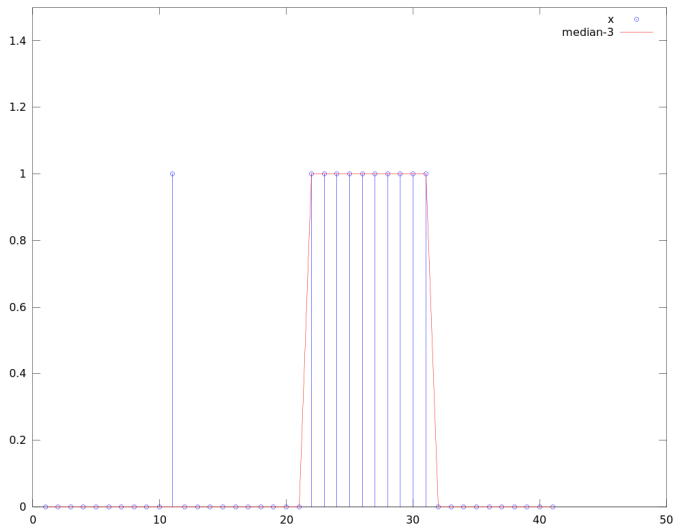
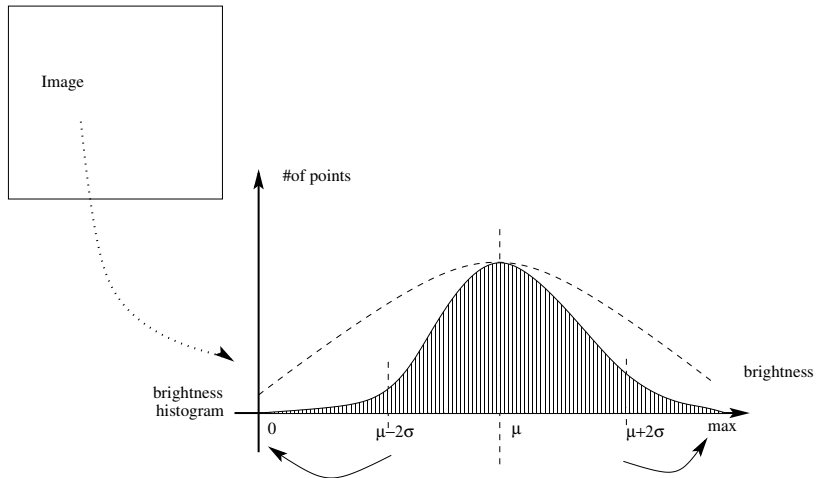


Image brightness normalization



Other (non-image) uses

- ▶ WERA array radar: spatial freq = direction, time freq = Doppler
- ▶ Smart antenna: spatial filter = selection of a user direction

Prepare for the lab!

- ▶ Understanding frequency in 2D
- ▶ Calculate a simple 2D Fourier transform: practice on 2x2 or 4x4

pictures:

0	0	0	0
1	1	1	1
0	0	0	0
1	1	1	1

0	1	0	1
0	1	0	1
0	1	0	1
0	1	0	1

1	0	1	0
0	1	0	1
1	0	1	0
0	1	0	1

- ▶ Apply a linear filter to above pictures.
- ▶ Understand a median. Why it is a nonlinear operation?