

## Spectrum Analysis

Dr Yvan Petillot

## Section Contents

- Introduction
- Periodogram
- Correlogram
- Windowing and spectrum analysis

## Spectrum analysis

Detection of signals (radar)

Signal modeling

Noise removal

Basic idea: *Determine the correlation between elements of a signal to enable modeling, compression. Determine power spectral content of the signal to enable noise removal, detection,...*

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Spectrum analysis

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## Correlation

- Signal Modeling
- Related to statistical description of signals (noise)
- Related to the Power Spectral Density (PSD)

$$R_{x_1x_1}(n) = x_1(n) \otimes x_1(n) = \sum_{m=-\infty}^{\infty} x_1(m)x_1(n+m) \quad \text{Autocorrelation}$$

↓ DFT

$$S_N(k) = \frac{1}{N} X(k)X^*(k) = \frac{1}{N} |X(k)|^2 \quad \text{Power spectral density}$$

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Spectrum analysis

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## Spectrum analysis

Why spectrum analysis?

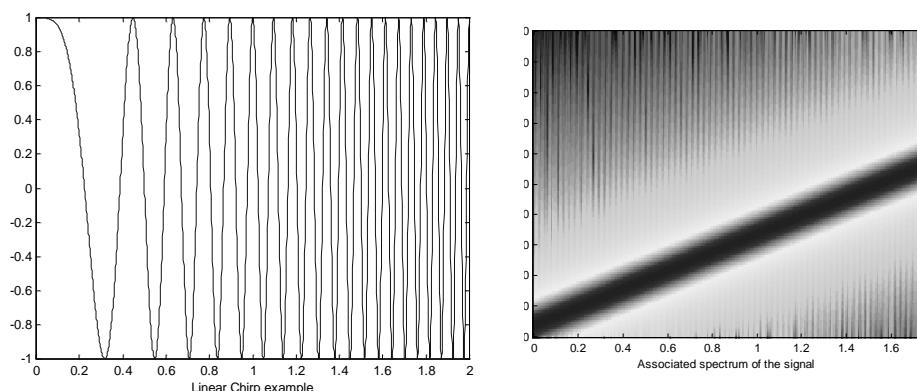
Filtering is a convolution in spectral domain

Information representation in the spectral domain can be:

- Easier to interpret
- Easier to visualize
- More concentrated
- More natural (audio frequencies)

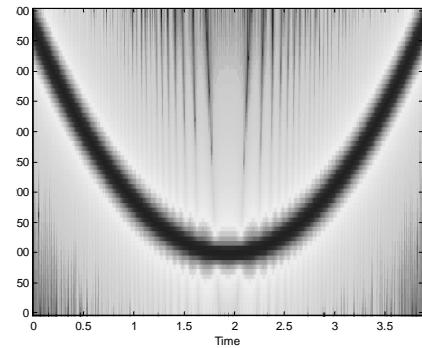
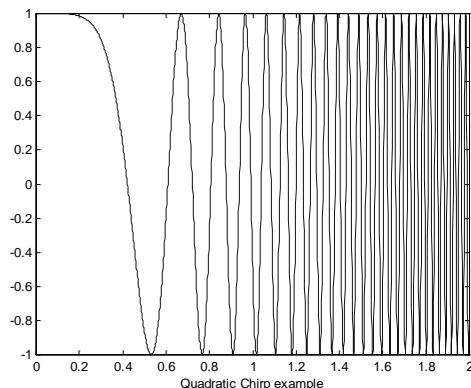
## Spectrum analysis

### Example



## Spectrum analysis

### Example

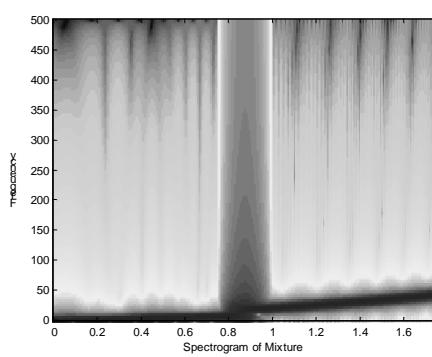
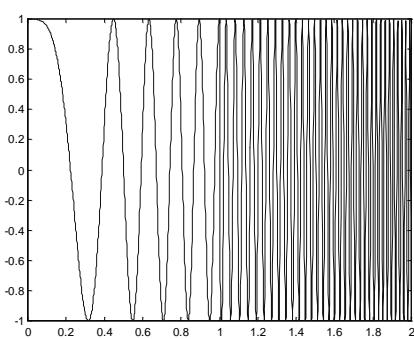


Spectrum analysis

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## Spectrum analysis

### Example

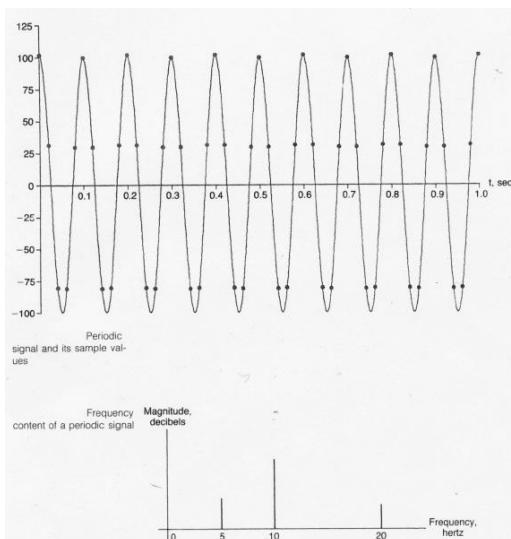


Spectrum analysis

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## Spectrum analysis

Example:



Spectrum analysis

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## Periodogram based spectral estimation

Technique:

$$S_N(k) = \frac{1}{N} |X(k)|^2 = \frac{1}{N} X(k)X^*(k), \quad k = 0, 1, \dots, N-1$$

Calculated using DFT.

$$x_1(n) \rightarrow \boxed{\text{DFT}} \rightarrow \left| \cdot \right|^2 \rightarrow S_N(k) = \frac{1}{N} |X(k)|^2$$

Problem?

What about long signals (thousands of samples)?

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## Periodogram based spectral estimation

Solution:

$x(n)$  is decomposed in  $M$  smaller overlapping sections.

The spectrum is calculated for each section

$$S_N(k) = \frac{1}{M} \sum_{m=1}^M S_{Nm}(k)$$

Assumes stationarity and ergodicity

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## Correlogram

$$R_{x_1x_1}(n) = x_1(n) \otimes x_1(n) = \sum_{m=-\infty}^{\infty} x_1(m)x_1(n+m) \quad \text{Autocorrelation}$$

$$R_N(p) = \frac{1}{N} \sum_{m=0}^{N-1-p} x_1(m)x_1(m+p) \quad \text{Correlogram}$$

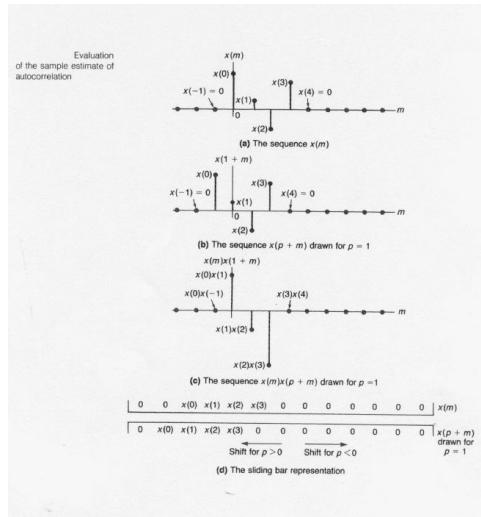
Same as autocorrelation with extra  $1/N$  term as summation done only on  $N$  terms

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## Correlogram



Spectrum analysis

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## Correlogram

Consider the z transform:

$$\begin{aligned} S_N(z) &= \frac{1}{N} X(z)X(z^{-1}) \\ &= \frac{1}{N} \sum_{n=0}^{N-1} x(n)z^{-n} \sum_{m=0}^{N-1} x(m)z^m \\ &= \frac{1}{N} \sum_{n=0}^{N-1} \sum_{m=0}^{N-1} x(n)x(m)z^{m-n} \end{aligned}$$

It can be shown that

$$\begin{aligned} S_N(z) &= \sum_{p=-(N-1)}^{N-1} r_N(p)z^{-p} \\ &= \sum_{p=0}^{N-1} r_N(p)[z^{-p} + z^p] - r_N(0) \end{aligned}$$

Therefore (see Notes):

$$S_N(k) = R_N(k) + R_N(k)^* - r_N(0)$$

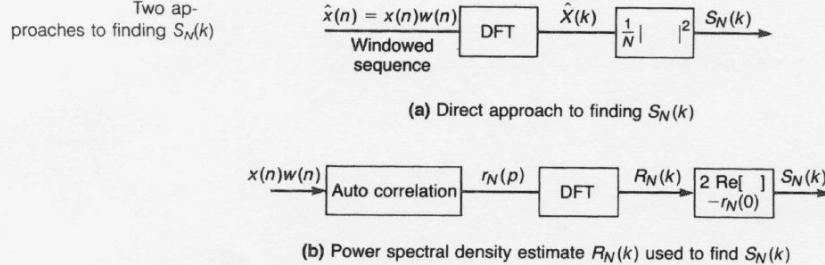
and

$$S_N(k) = 2R_e[R_N(k)] - r_N(0)$$

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## Periodogram versus Correlogram



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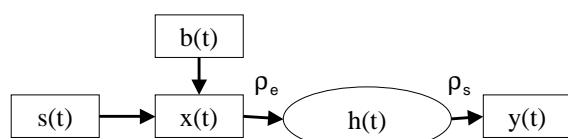
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## A part on power density

Why use power spectral density?

$$\rho = \text{SNR} = \frac{P_{\text{signal}}}{P_{\text{bruit}}} = \frac{\int S(v)dv}{\int B(v)dv} = \frac{\sum_{k=0}^{N-1} S_N(k)}{\sum_{k=0}^{N-1} B_N(k)}$$

In a linear time invariant system (filter)



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## A part on power density

Matched filtering: optimizes SNR out  $\rho_s$

$$h(v) = e^{-j2\pi v T} \frac{s^*(v)}{B(v)}$$

White noise case:

$$h(v) = e^{-j2\pi v T} s^*(v)$$

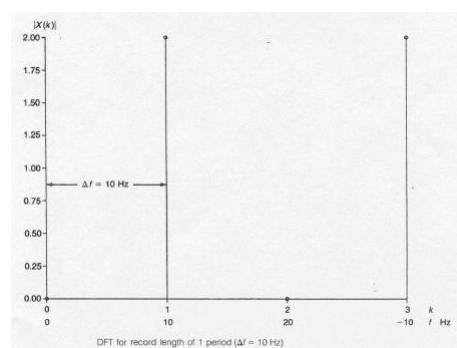
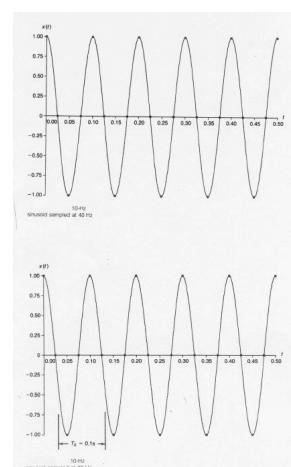
$$h(t) = s(T-t) \quad \text{correlation}$$

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## Windows for spectrum analysis

Experiment:



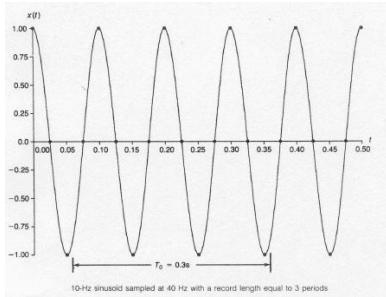
10 Hz signal sampled at 40 Hz analysed using a 0.1 s window

Spectrum analysis

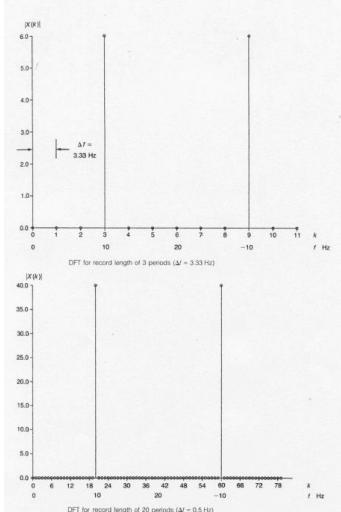
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## Windows for spectrum analysis

### Experiment:



10 Hz signal sampled at 40 Hz analysed using a 0.3 / 2 s window



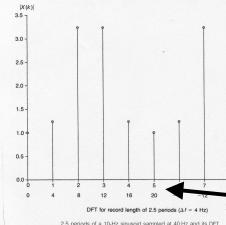
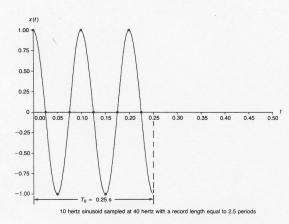
DFT Record Length of 1 period ( $\Delta f = 3.33 \text{ Hz} / \Delta f = 0.5 \text{ Hz}$ )

Spectrum analysis

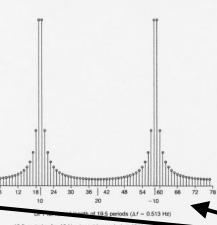
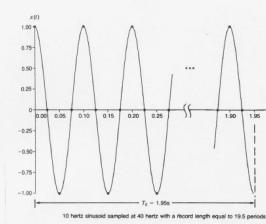
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## Windows for spectrum analysis

### Experiment: 10 Hz signal sampled at 40 Hz analysed using a 0.25 / 1.95s window



DFT Record Length of 1 period ( $\Delta f = 4 \text{ Hz} / \Delta f = 0.513 \text{ Hz}$ )



Check frequency content?

Check Levels

What is happening?

Leakage  
or  
Gibbs phenomenon

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## Windows for spectrum analysis

Solution?

First clue: think about what a Discrete Fourier domain means in the time domain...

Second clue:

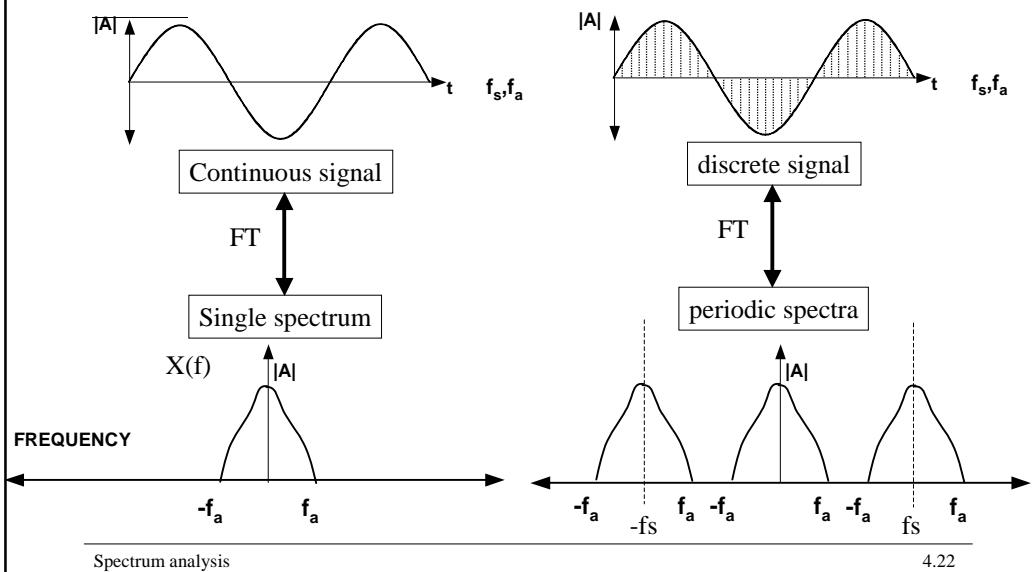
If you consider a integer number of periods what happens?

\_\_ What if it is a rational number of periods?

Spectrum analysis

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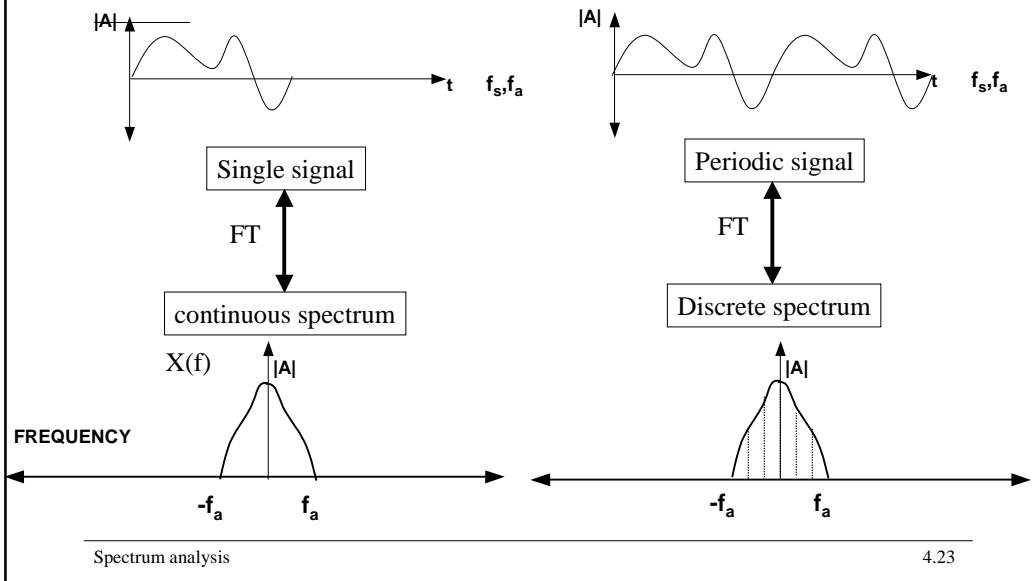
## Reminder: Discreteness and periodicity



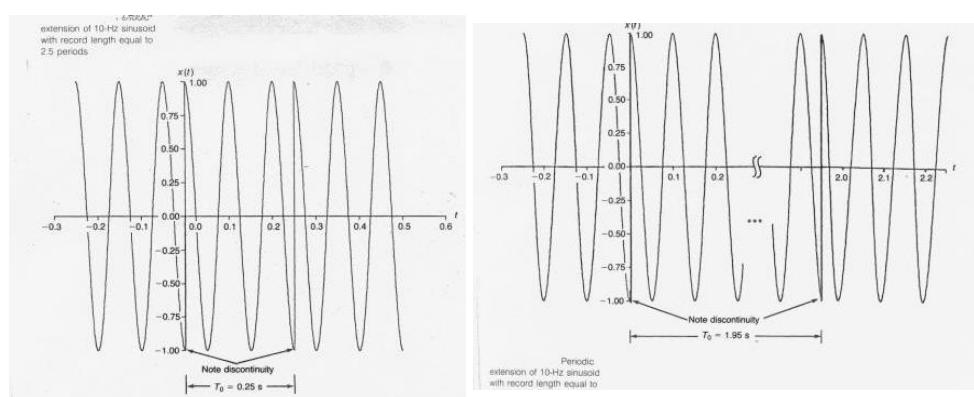
Spectrum analysis

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## Reminder: Discreteness and periodicity



## Windows for spectrum analysis



Spectrum analysis

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## Windows for spectrum analysis

### Case of real signals:

non-periodic

leakage is bound to happen unless the signal is the same on both side of the window of analysis:

*Will not happen naturally for all signals and all windows*

*Will restrict the choice of windows / frequency resolution*

*Is unpractical for averaged periodogram estimation*

## Windows for spectrum analysis

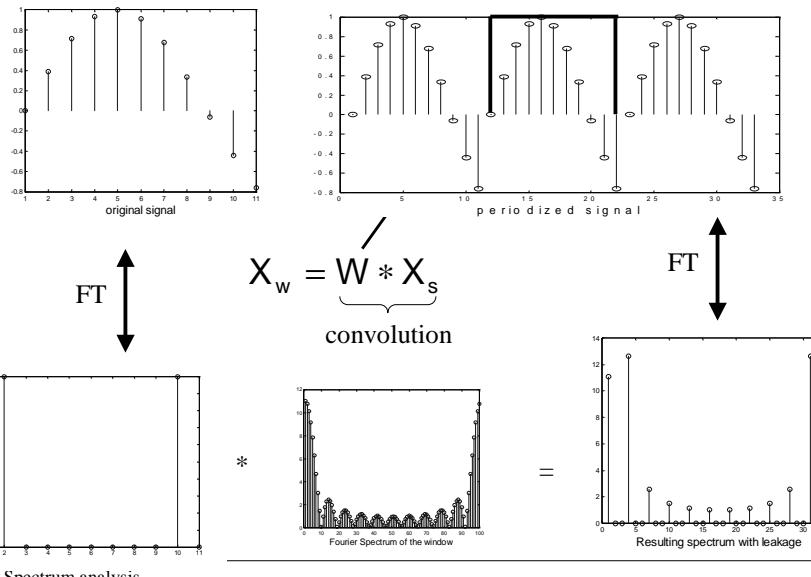
Solution is windowing the signals so that each end of the window have the same value!

Effect of windowing?

Theory:

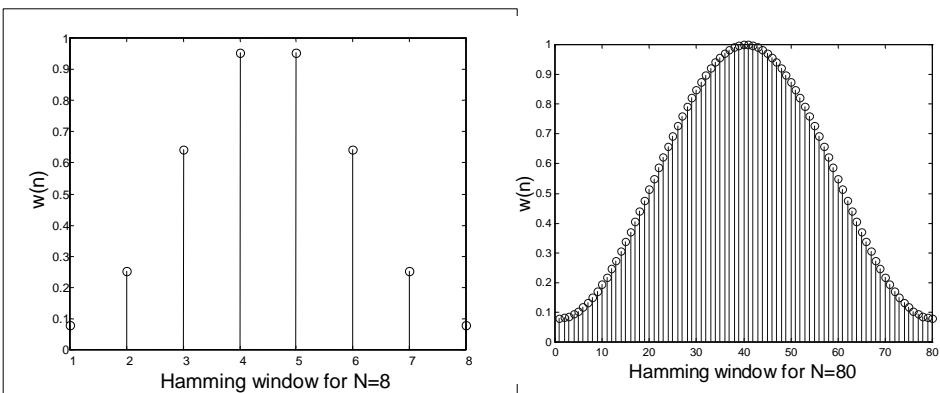
$$\begin{aligned}x_s &= x(t) \sum_{k=0}^{\infty} \delta(t - kT) & X_w &= W X_s \\X_s &= \frac{1}{T_s} \sum_{k=0}^{\infty} X(f - kf_s) & X_w &= \underbrace{W * X_s}_{\text{convolution}}\end{aligned}$$

## Square window



## Optimised classical windows

$$\underline{\text{Hamming:}} \quad w(n) = 0.54 - 0.46 * \cos\left(\frac{2\pi n}{N-1}\right)$$

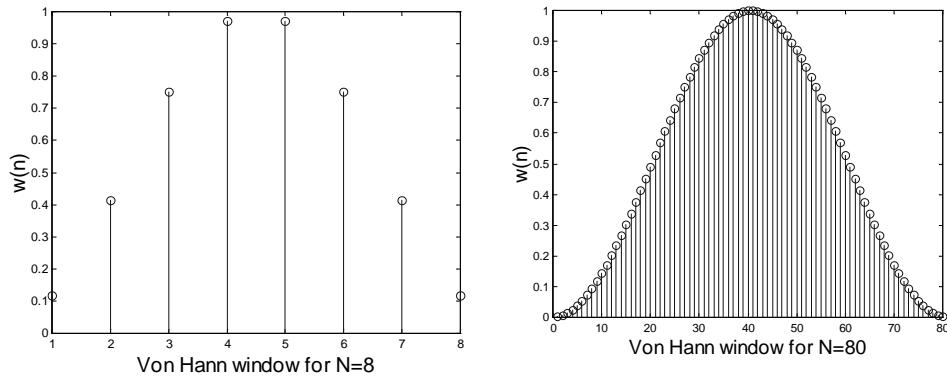


Spectrum analysis

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## Optimised classical windows

Hanning:  $w(n) = 0.50 - 0.50 * \cos\left(\frac{2\pi n}{N-1}\right)$



Spectrum analysis

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## Optimised classical windows

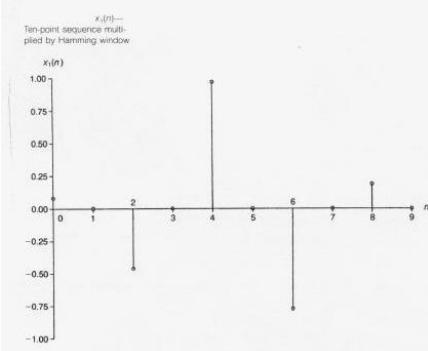
### Difference between windows ?

Window type	Mathematical expression	Sidelobes (dB)	Transition width	Stop band attenuation
Rectangular	$w(n) = 1, \quad 0 \leq n \leq N - 1$	-13	$0.9/N$	-21
Hamming	$w(n) = 0.54 - 0.46 \cos\left(\frac{2\pi n}{N-1}\right) \quad 0 \leq n \leq N-1$	-31	$3.1/N$	-44
Hanning	$w(n) = 0.50 - 0.40 \cos\left(\frac{2\pi n}{N-1}\right) \quad 0 \leq n \leq N-1$	-41	$3.3/N$	-53
Blackman	$w(n) = 0.42 - 0.50 \cos\left(\frac{2\pi n}{N-1}\right) + 0.08 \cos\left(\frac{4\pi n}{N-1}\right) \quad 0 \leq n \leq N-1$	-57	$5.5/N$	-74

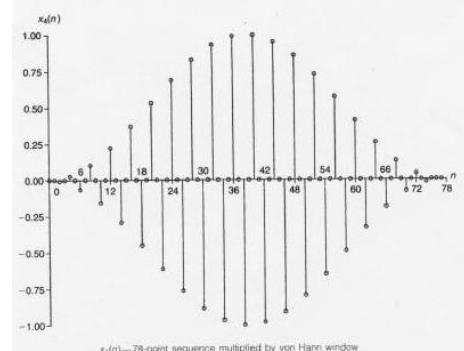
Spectrum analysis

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## Example



10 points sequence multiplied by hamming window

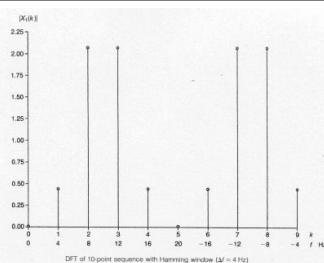


75 points sequence multiplied by hanning window

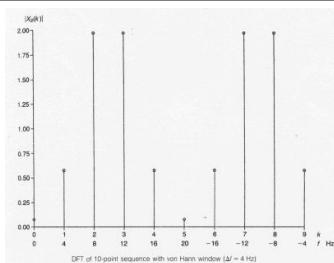
Spectrum analysis

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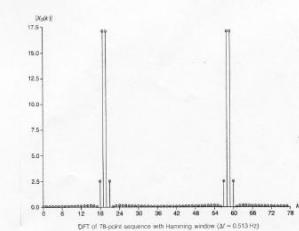
## Example



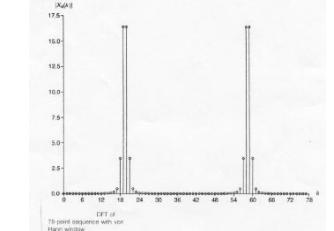
10 points DFT for Hamming window ( $\Delta f = 4 \text{ Hz}$ )



10 points DFT for Hanning window ( $\Delta f = 4 \text{ Hz}$ )



75 points DFT for Hamming window ( $\Delta f = 0.513 \text{ Hz}$ )



75 points DFT for Hanning window ( $\Delta f = 0.513 \text{ Hz}$ )

Spectrum analysis

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