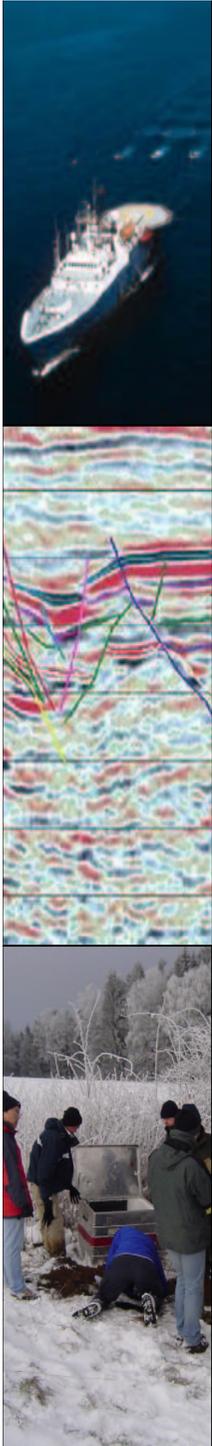


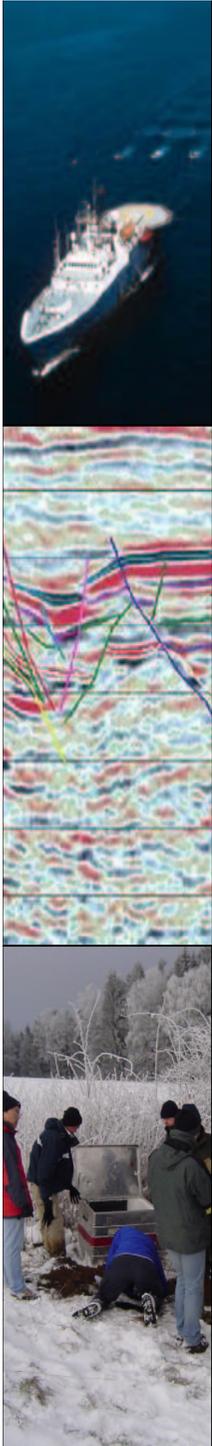
Geophysical Data Processing

- **Digitization**
 - Dynamic range
- **Spectral analysis**
 - Fourier analysis
 - Sampling
 - Spectra in space and time domain
- **Waveform processing**
 - Convolution
 - Deconvolution
 - Correlation
 - Digital Filtering

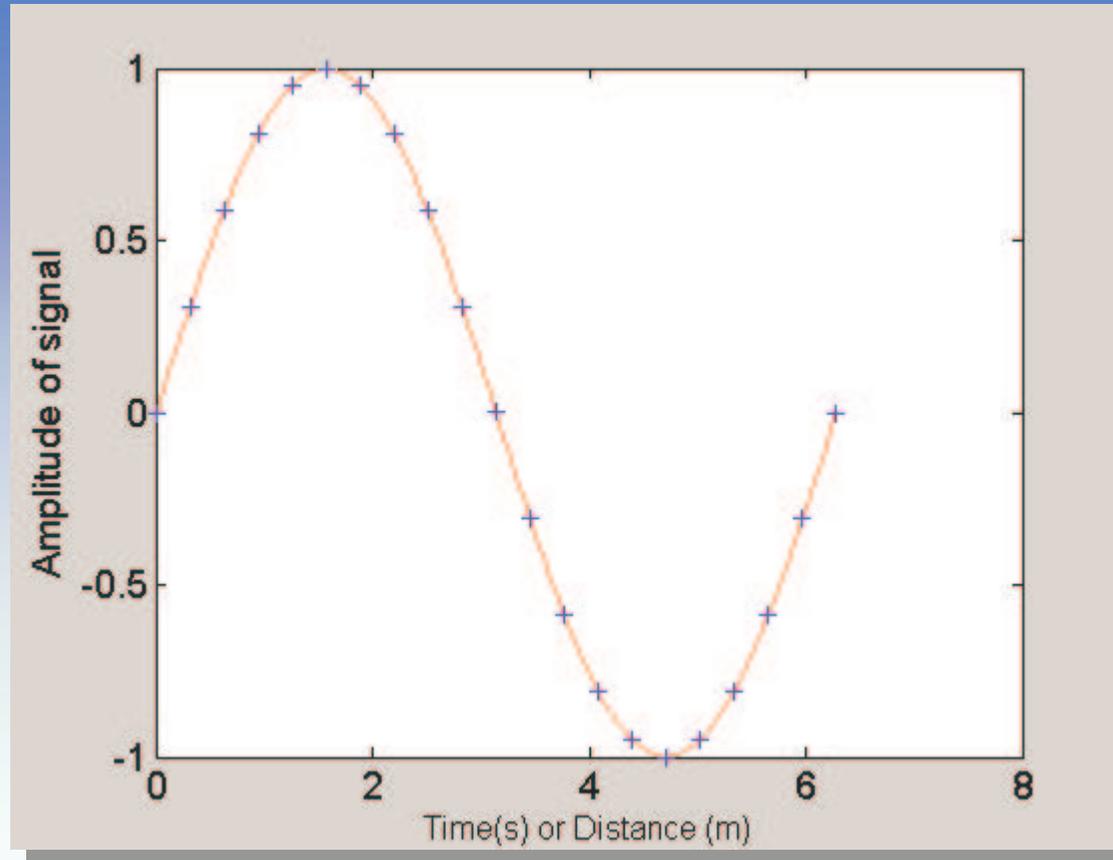


Digitization

- What happens when I **digitize** a signal (ground motion, temperature, etc.) in **space** and/or **time**?
- What are the consequences of a particular **sampling rate** on the information content?
- How can we **treat (process, transform)** the observed signals to extract relevant information?



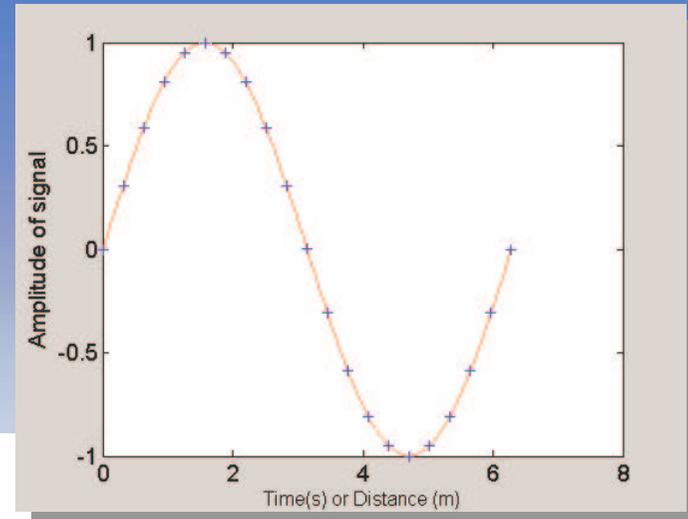
Digitization



Analog and digital (+) representation of a sinusoidal function

Wavelength, period, etc.

The most important concepts you will ever need in processing observations are **spatial and temporal frequencies**



T period
 f frequency
 ω angular frequency

$$T=1/f$$
$$\omega=2\pi f$$

temporal frequencies

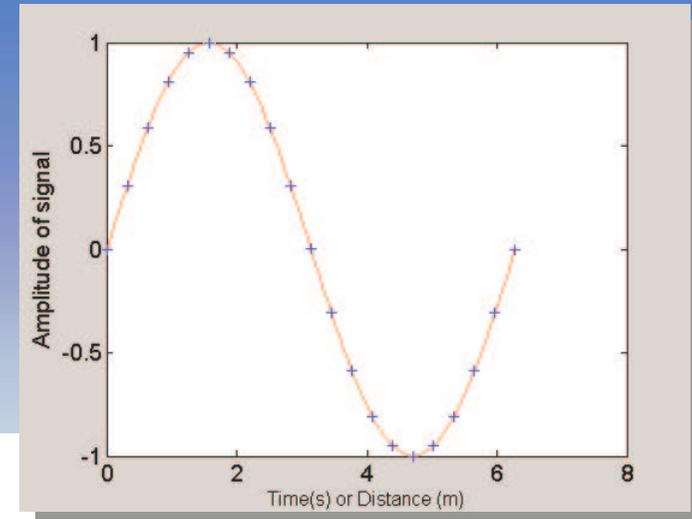
Harmonic oscillation in time:

$$f(t) = A \sin(\omega t) = A \sin(2\pi f t) = A \sin\left(\frac{2\pi}{T} t\right)$$

A motion amplitude

Wavelength, period, etc.

... and the space analogue ...



λ wavelength
 k spatial wavenumber

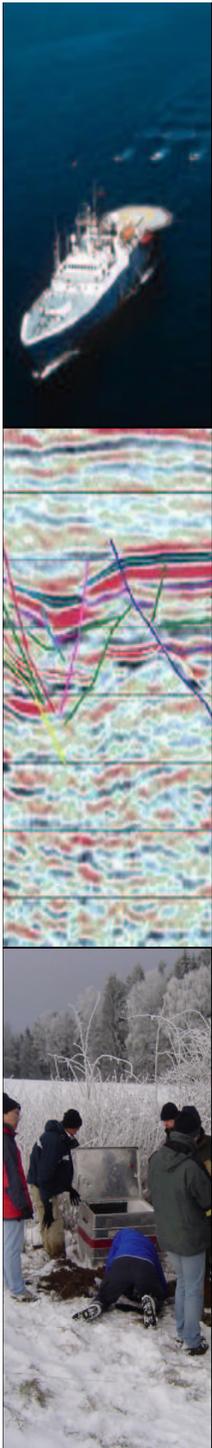
$$k = 2\pi/\lambda$$

spatial frequencies

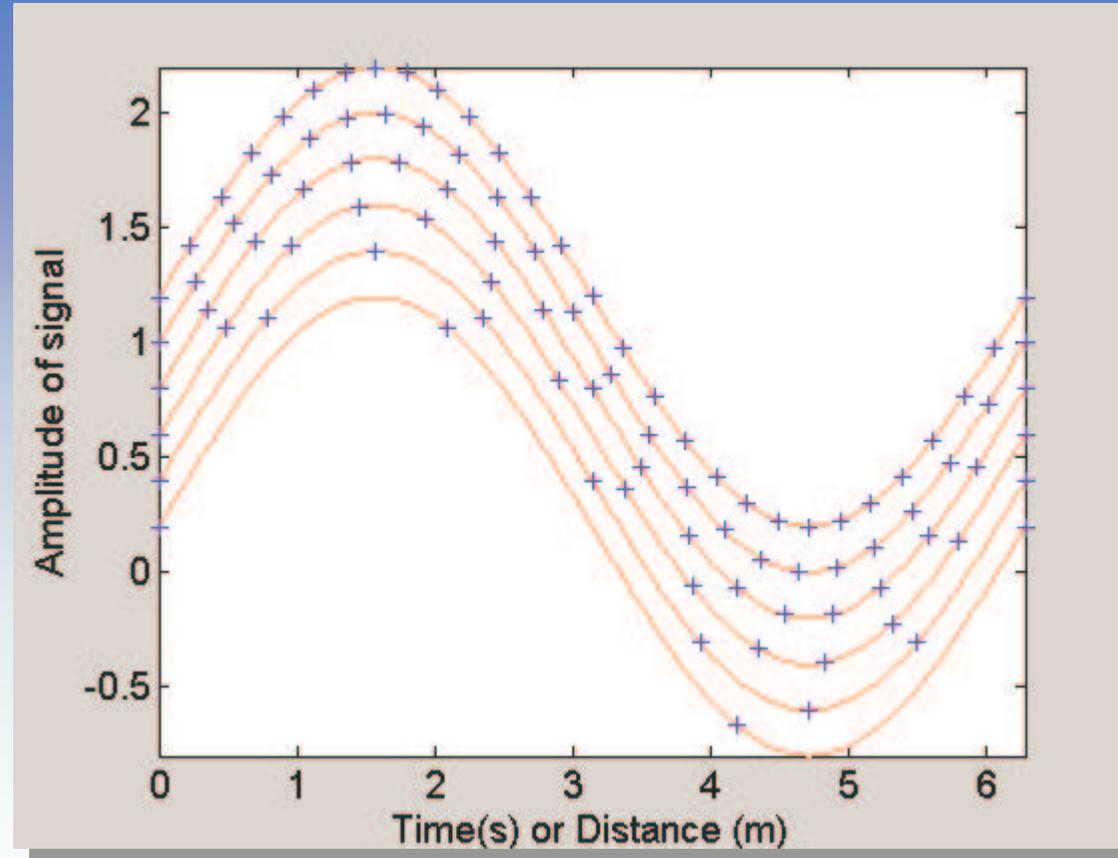
Harmonic oscillation in time:

$$f(x) = A \sin(kx) = A \sin((2\pi/\lambda) x)$$

A motion amplitude



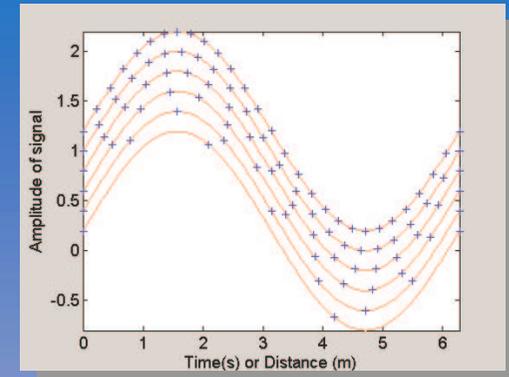
Sampling rate



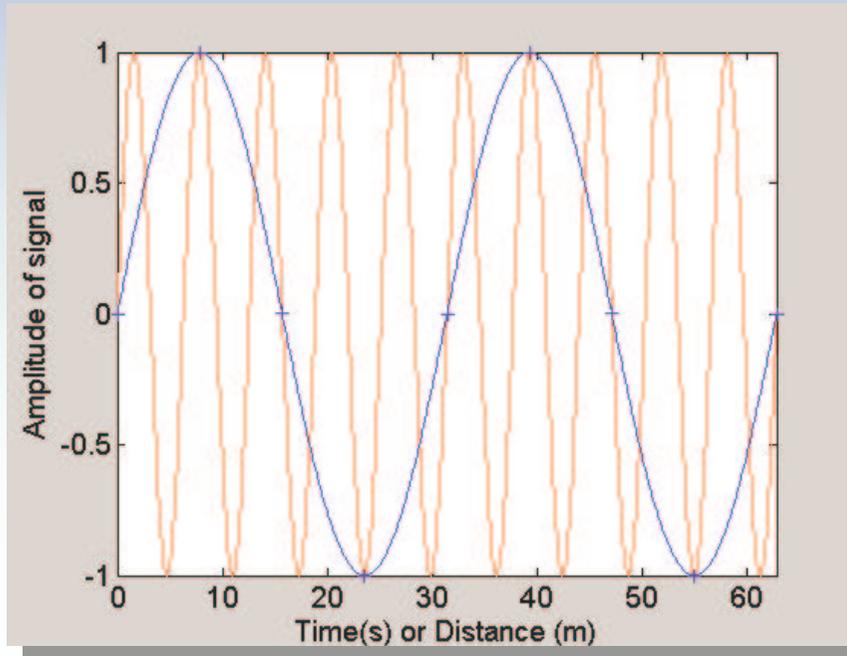
Sampling frequency, sampling rate is the number of sampling points per unit distance or unit time. Examples?



Nyquist Frequency (Wavenumber, Interval)

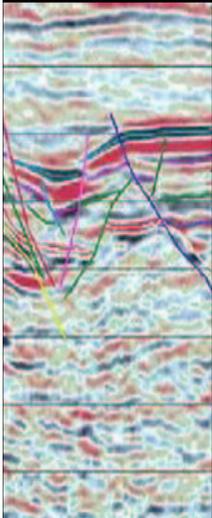


The frequency **half of the sampling rate dt** is called the **Nyquist frequency $f_N=1/(2dt)$** . The distortion of a physical signal higher than the Nyquist frequency is called **aliasing**.

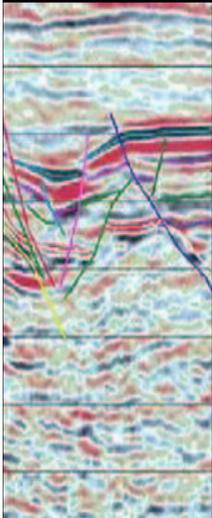


The frequency of the **physical signal** is $> f_N$ is sampled with (+) leading to the erroneous **blue** oscillation.

What happens in space?
How can we avoid aliasing?



A cattle grid



Dynamic range

What is the precision of the sampling of our physical signal in amplitude?

Dynamic range: the ratio between largest measurable amplitude A_{\max} to the smallest measurable amplitude A_{\min} .

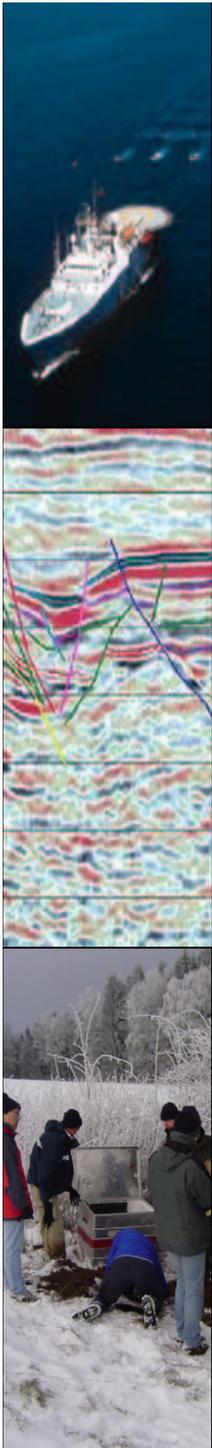
The unit is Decibel (dB) and is defined as the ratio of two power values (and power is proportional to amplitude square)

In terms of amplitudes

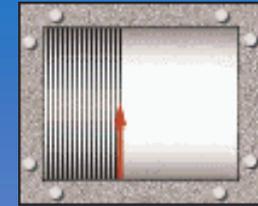
$$\text{Dynamic range} = 20 \log_{10}(A_{\max}/A_{\min}) \text{ dB}$$

Example: with 1024 units of amplitude ($A_{\min}=1$, $A_{\max}=1024$)

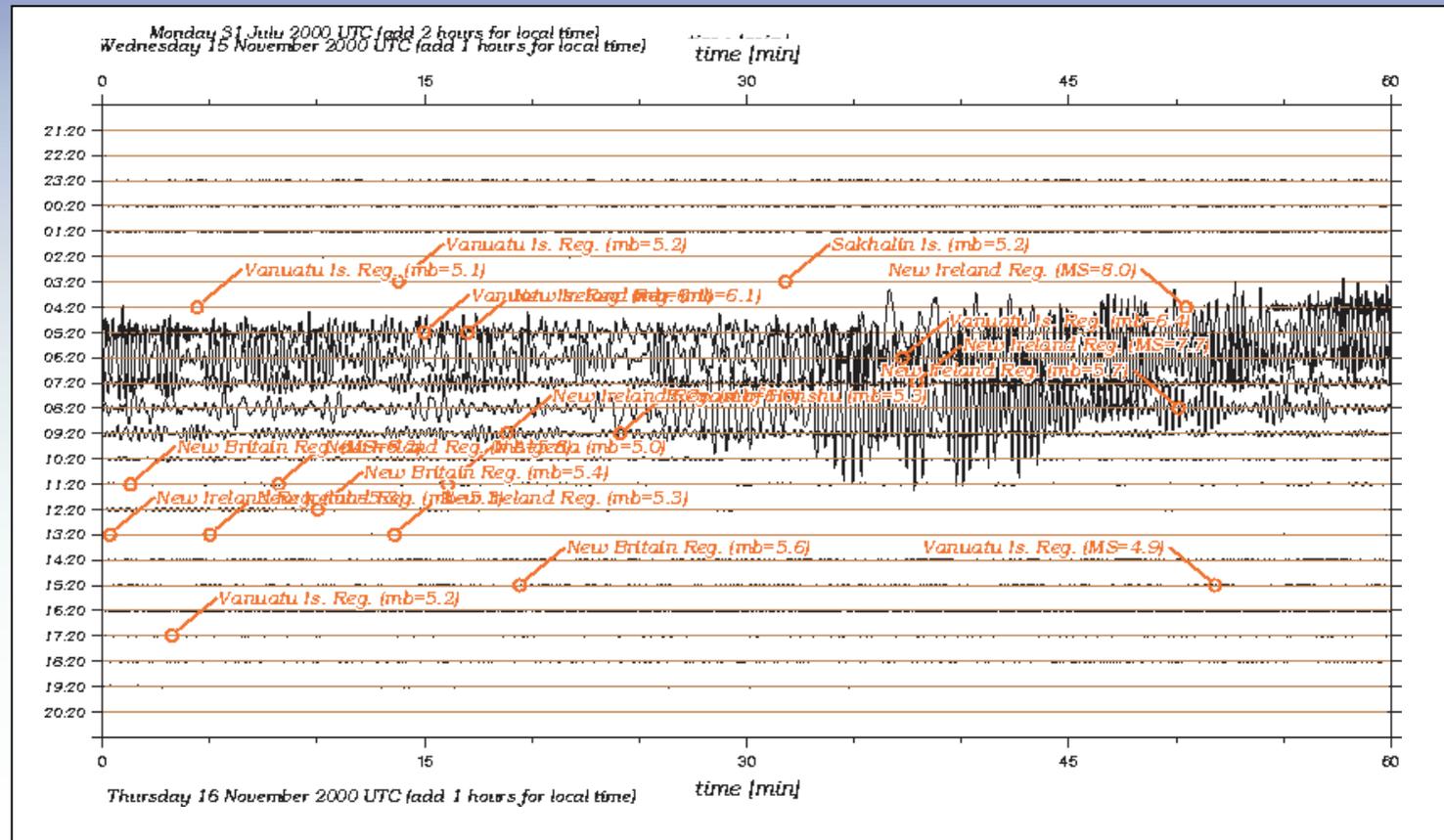
$$20 \log_{10}(1024/1) \text{ dB} \approx 60 \text{ dB}$$



Signal and Noise



Almost all signals contain **noise**. The **signal-to-noise ratio** is an important concept to consider in all geophysical experiments. Can you give examples of noise in the various methods?

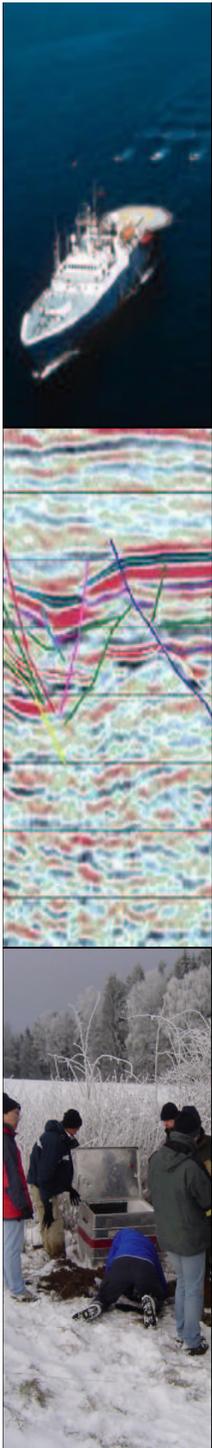


Spectral Analysis

Spectral analysis is so central to all natural (and other) sciences that its role can not be **overemphasized!**

With **spectral analysis** we can get answers to questions like:

- What (spatial or temporal) frequencies are contained in my signal?
- Is there a **periodic signal** in my observations?
- Do I have to account for the **response** of my measurement instrument (e.g. seismometer) to obtain the physical signal?
- Do I have to **filter** the observations first before seeing the physical signal?
- and, and, and ...

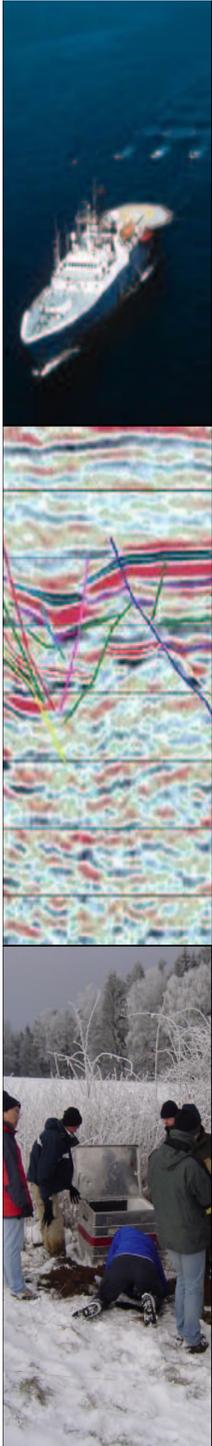


Harmonic Analysis - Spectral Synthesis

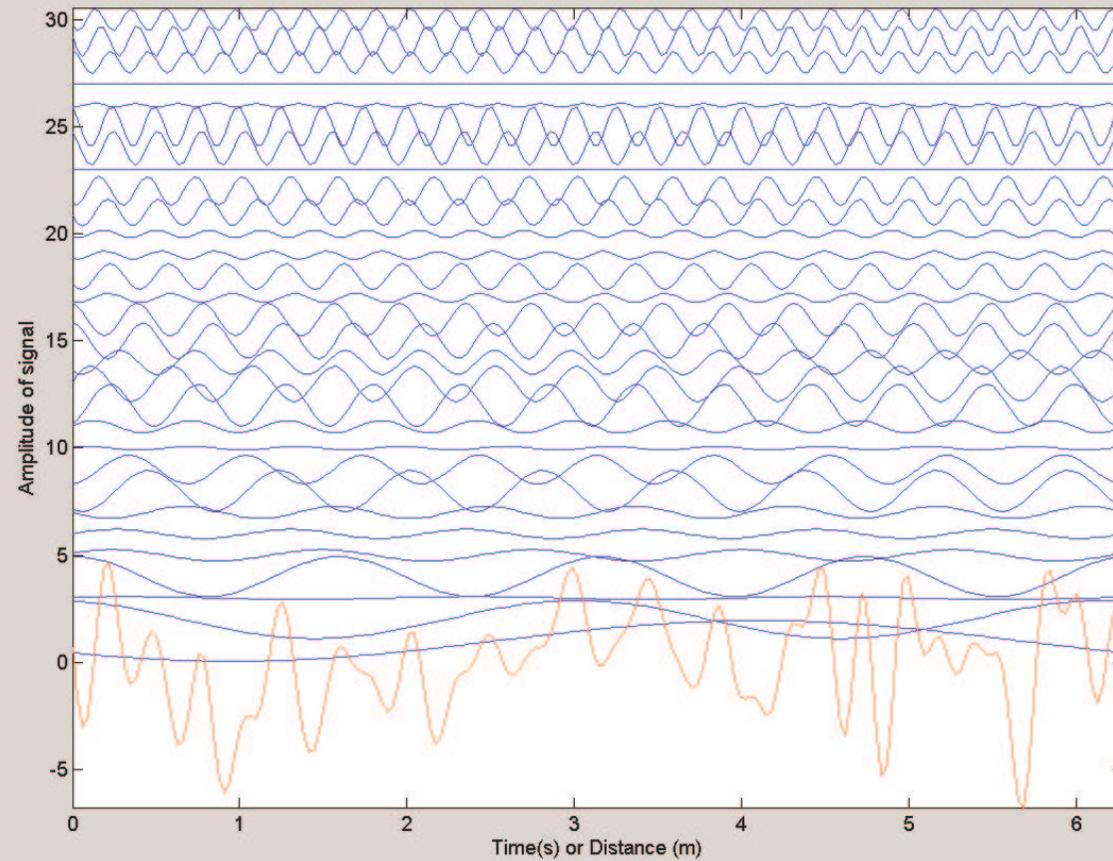
At the heart of **spectral analysis** is an extremely powerful concept, that is one of the most important theorems in mathematical physics:

Any arbitrary signal can be obtained by superposition of harmonic (sinusoidal) signals.

Furthermore: the representation of physical systems in **time and space** or in **frequency and wavenumber domain** is equivalent! There is no loss of information when going from one space to the other and back.



Spectral synthesis



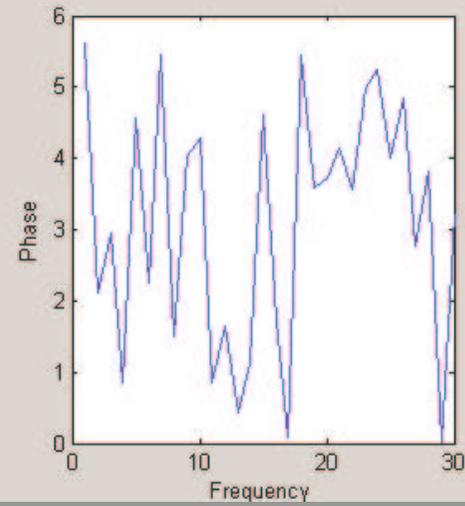
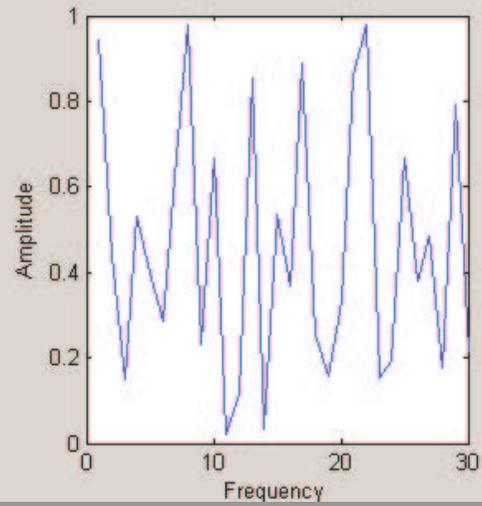
The **red** trace is the sum of all **blue** traces!

The spectrum

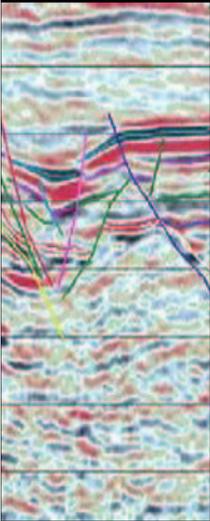
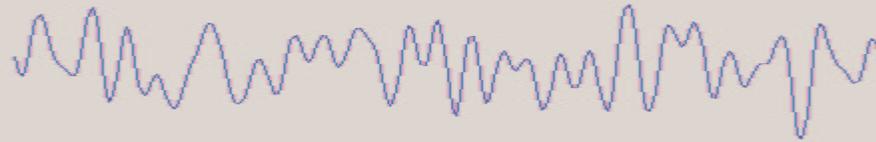
Amplitude spectrum

Phase spectrum

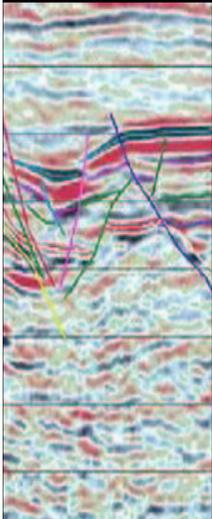
Fourier space

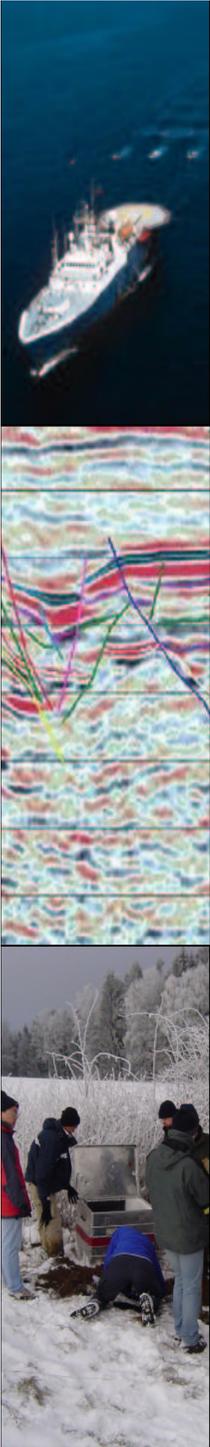


Physical space



Fourier decomposer





Let's get formal!

A sinusoidal function is represented as (a amplitude, λ wavelength):

$$y = a \sin\left(\frac{2\pi x}{\lambda}\right)$$

Ignoring phase shifts an arbitrary signal (a_0 at both ends) can be obtained by superposition of

$$f(x) = a_0 + \sum_n a_n \sin\left(2\pi x \frac{n}{2L}\right) \quad n = 1, \infty$$

Here L is the length of the space (or time) domain. Note the sequence of wavelengths (or periods): $2L, L, 2/3L, L/2 \dots$

The Fourier components

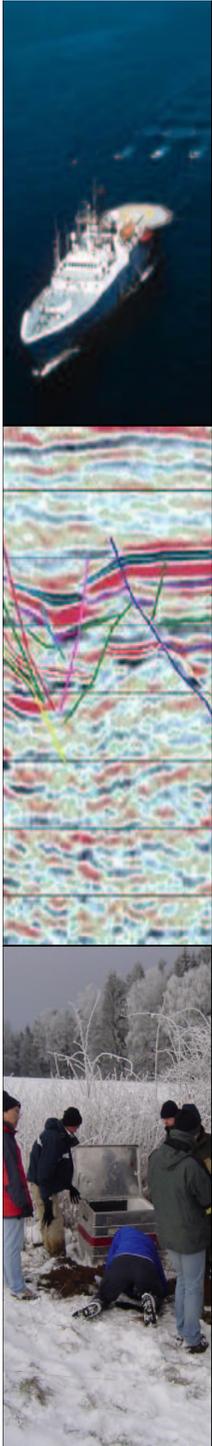
The amplitudes (a_n) of the signal (**Fourier components**) can be obtained by integrating over the signal:

$$a_0 = \frac{1}{L} \int_0^L f(x) dx$$

Average value of signal

$$a_n = \frac{2}{L} \int_0^L f(x) \sin \frac{n\pi x}{L} dx$$

Spectral component



Fourier: Space and Time

Space

x	space variable
L	spatial wavelength
$k=2\pi/\lambda$	spatial wavenumber
$F(k)$	wavenumber spectrum

Time

t	Time variable
T	period
f	frequency
$\omega=2\pi f$	angular frequency

Fourier integrals

With the complex representation of sinusoidal functions e^{ikx} (or $e^{i\omega t}$) the Fourier transformation can be written as:

$$f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} F(k) e^{-ikx} dx$$

$$F(k) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{ikx} dx$$

The Fourier Transform discrete vs. continuous

Whatever we do on the computer with data will be based on the discrete Fourier transform

continuous

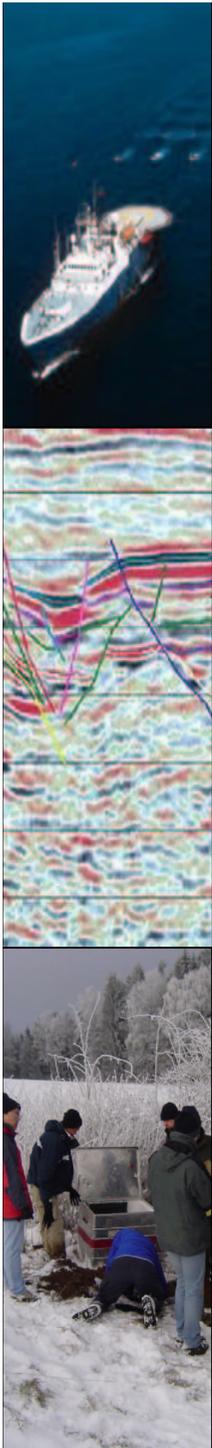
$$f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} F(k) e^{-ikx} dx$$

$$F(k) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{ikx} dx$$

discrete

$$F_k = \frac{1}{N} \sum_{j=0}^{N-1} f_j e^{-2\pi i k j / N}, k = 0, 1, \dots, N-1$$

$$f_k = \sum_{j=0}^{N-1} F_j e^{2\pi i k j / N}, k = 0, 1, \dots, N-1$$



The Fast Fourier Transform (FFT)

Most processing tools (e.g. octave, Matlab, Mathematica, Fortran, etc) have intrinsic functions for FFTs

Matlab FFT

```
>> help fft
```

FFT Discrete Fourier transform
FFT(X) is the discrete Fourier transform (DFT) of vector X. For matrices, the FFT operation is applied to each column. For N-D arrays, the FFT operation operates on the first non-singleton dimension.

FFT(X,N) is the N-point FFT, padded with zeros if X has less than N points and truncated if it has more.

FFT(X,[],DIM) or FFT(X,N,DIM) applies the FFT operation across the dimension DIM

For length N input vector x, the DFT is a length N vector X, with elements

$$X(k) = \sum_{n=1}^N x(n) \exp(-j \cdot 2 \cdot \pi \cdot (k-1) \cdot (n-1) / N), \quad 1 \leq k \leq N.$$

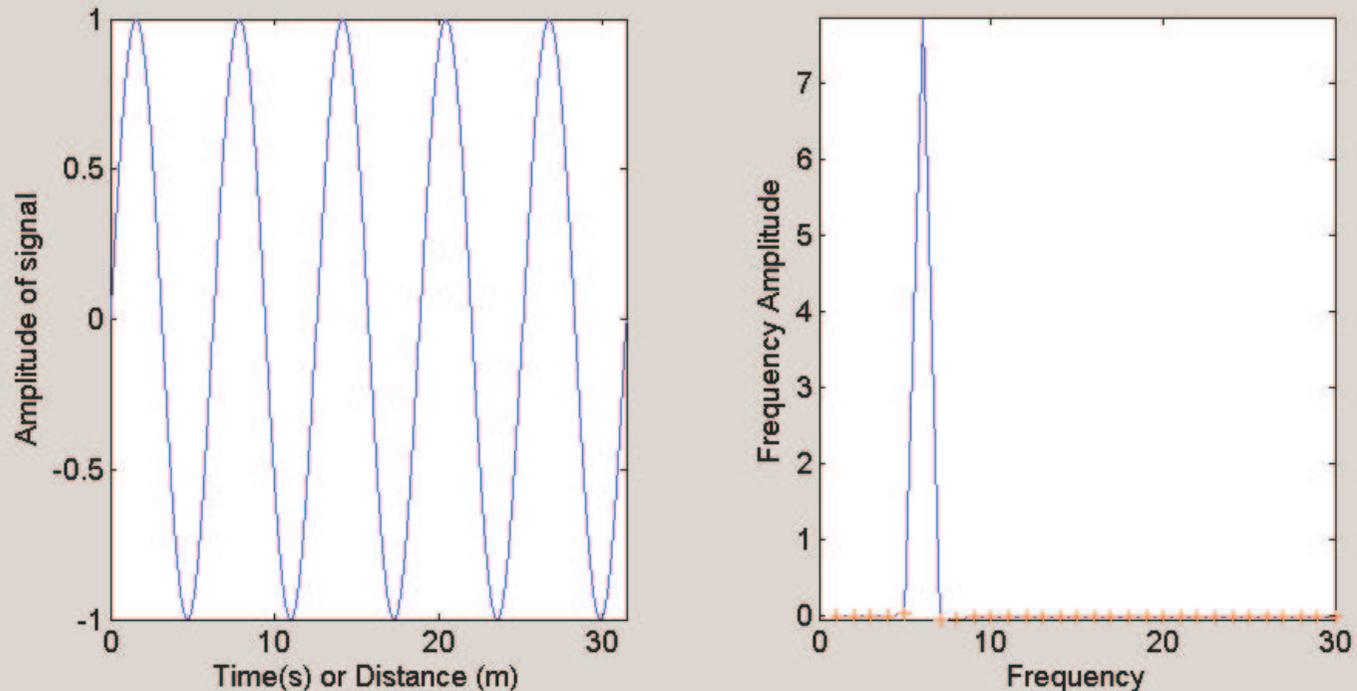
The inverse DFT (computed by IFFT) is given by

$$x(n) = (1/N) \sum_{k=1}^N X(k) \exp(j \cdot 2 \cdot \pi \cdot (k-1) \cdot (n-1) / N), \quad 1 \leq n \leq N.$$

See also IFFT, FFT2, IFFT2, FFTSHIFT.

Fourier Spectra: Main Cases

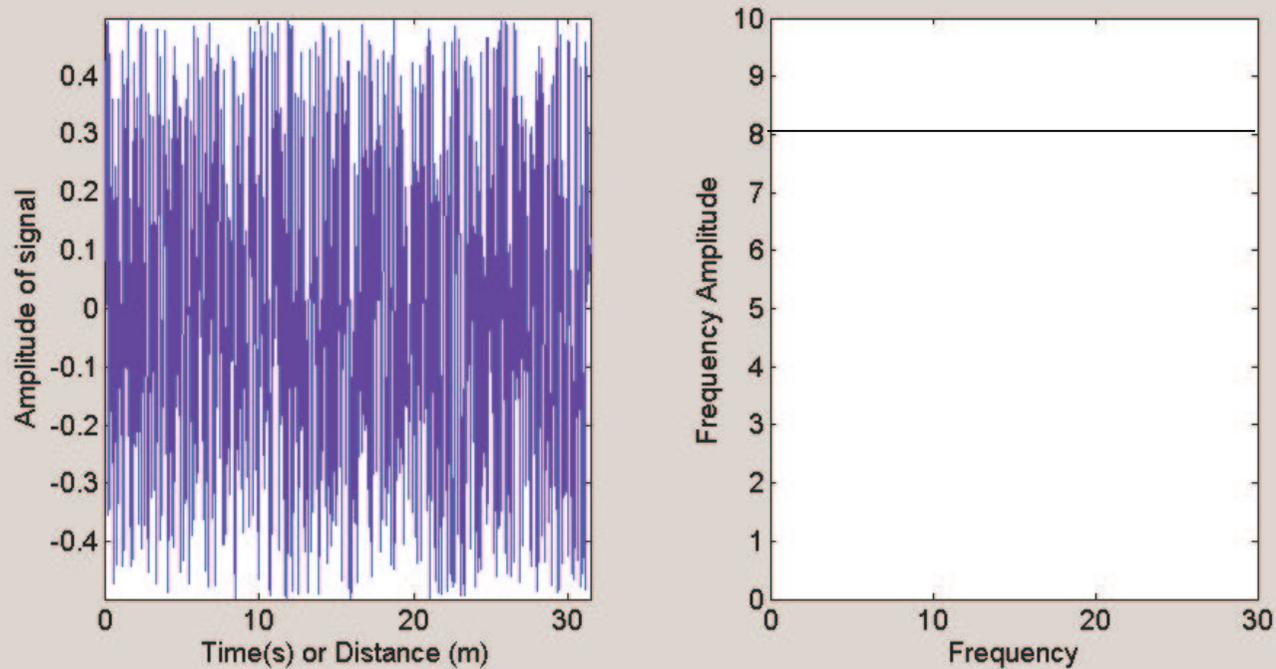
harmonic signals



The spectrum of a (monochromatic) harmonic signal in space (or time) is a spike („Delta function“) in the frequency domain.

Fourier Spectra: Main Cases

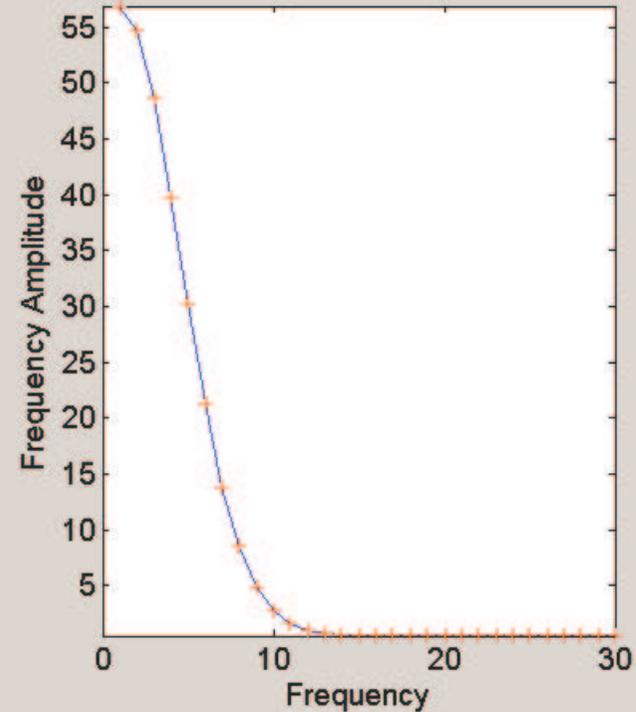
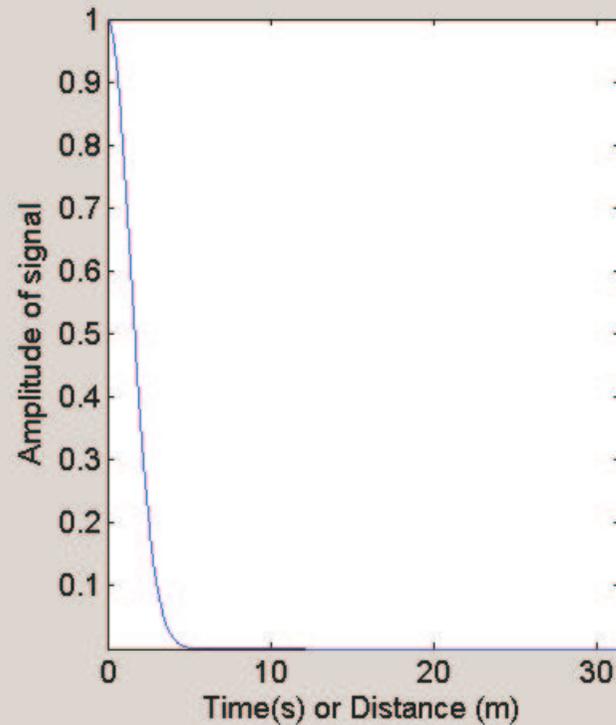
random signals



Random signals contain **all frequencies**. A spectrum with constant contribution of all frequencies is called a **white spectrum**

Fourier Spectra: Main Cases

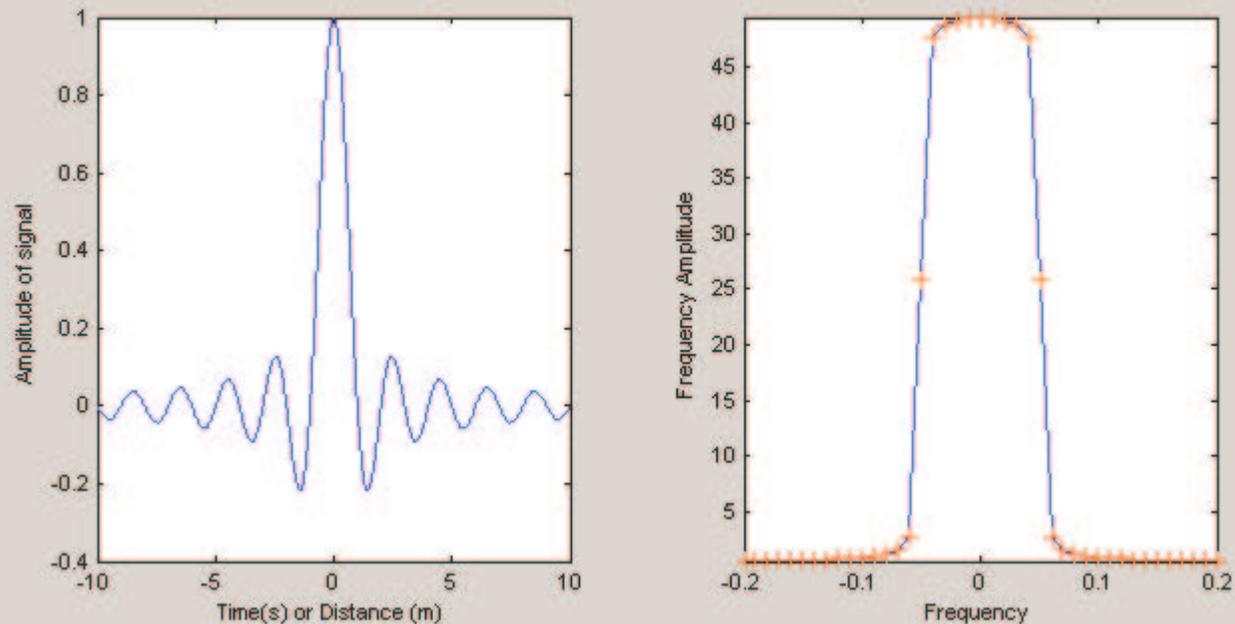
Gaussian signals



The spectrum of a Gaussian function will itself be a Gaussian function. How does the spectrum change, if I make the Gaussian narrower and narrower?

Fourier Spectra: Main Cases

Transient waveform



A **transient** wave form is a wave form limited in time (or space) in comparison with a harmonic wave form that is infinite

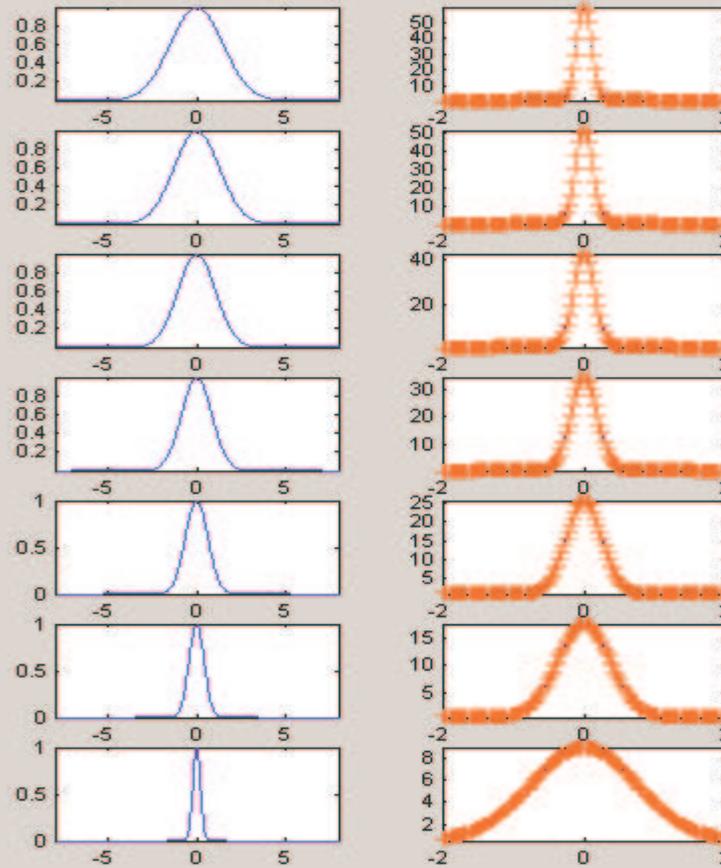
Puls-width and Frequency Bandwidth

time (space)

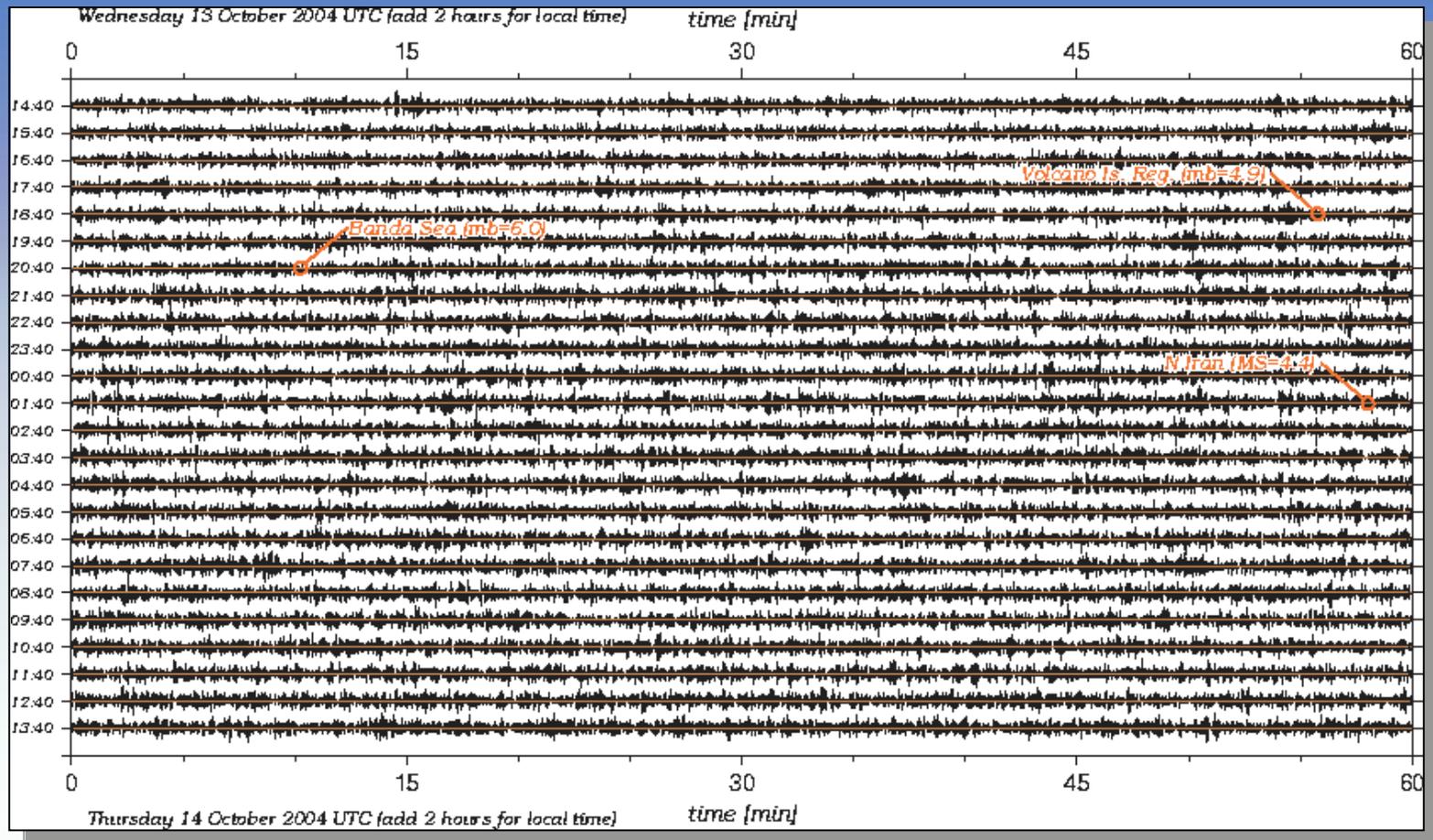
spectrum

Narrowing physical signal

Widening frequency band

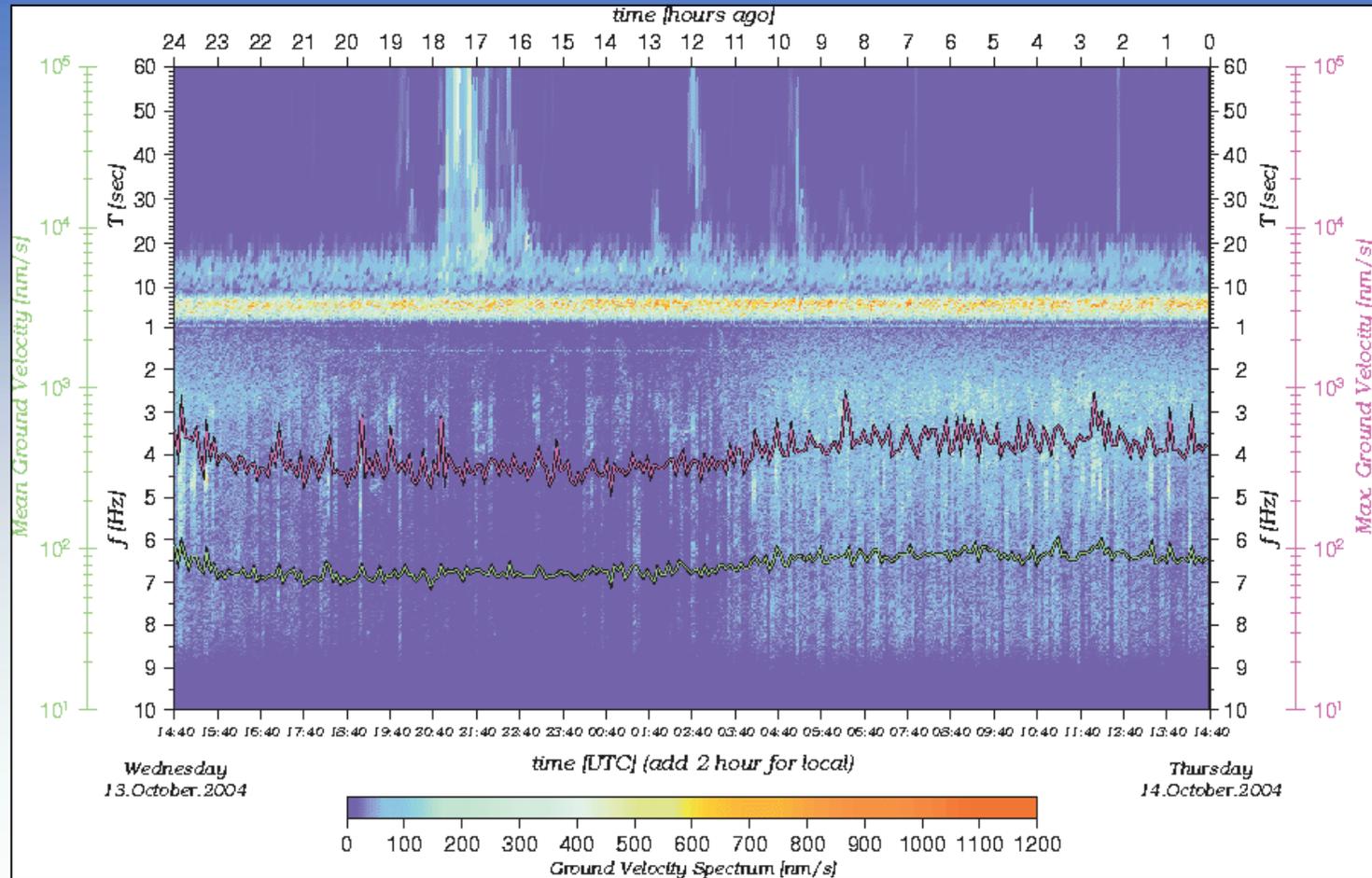


Spectral analysis: an Example



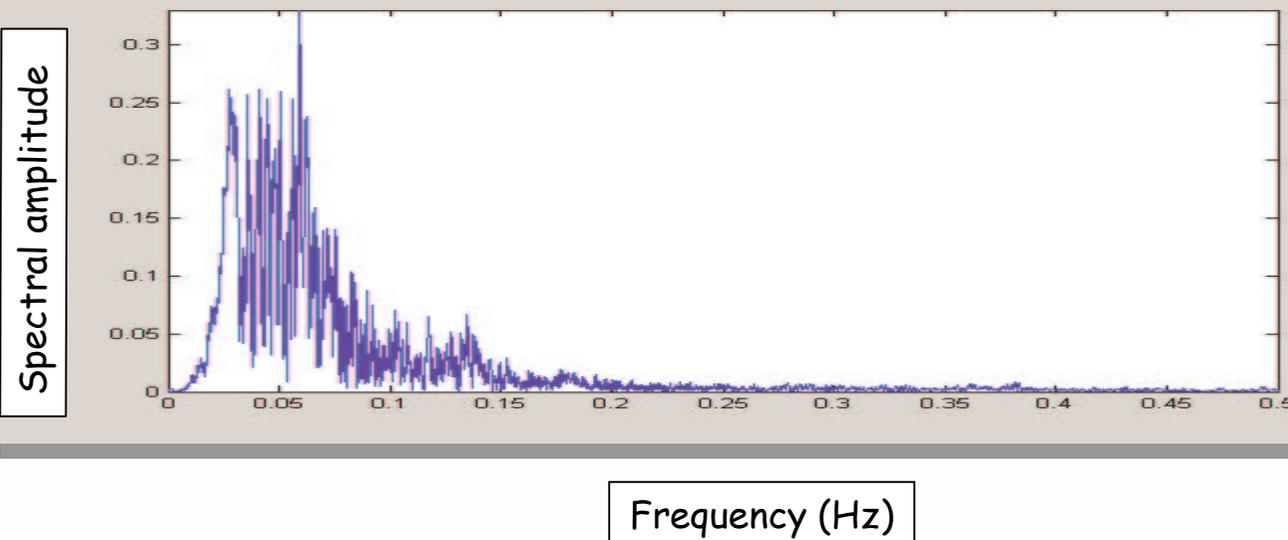
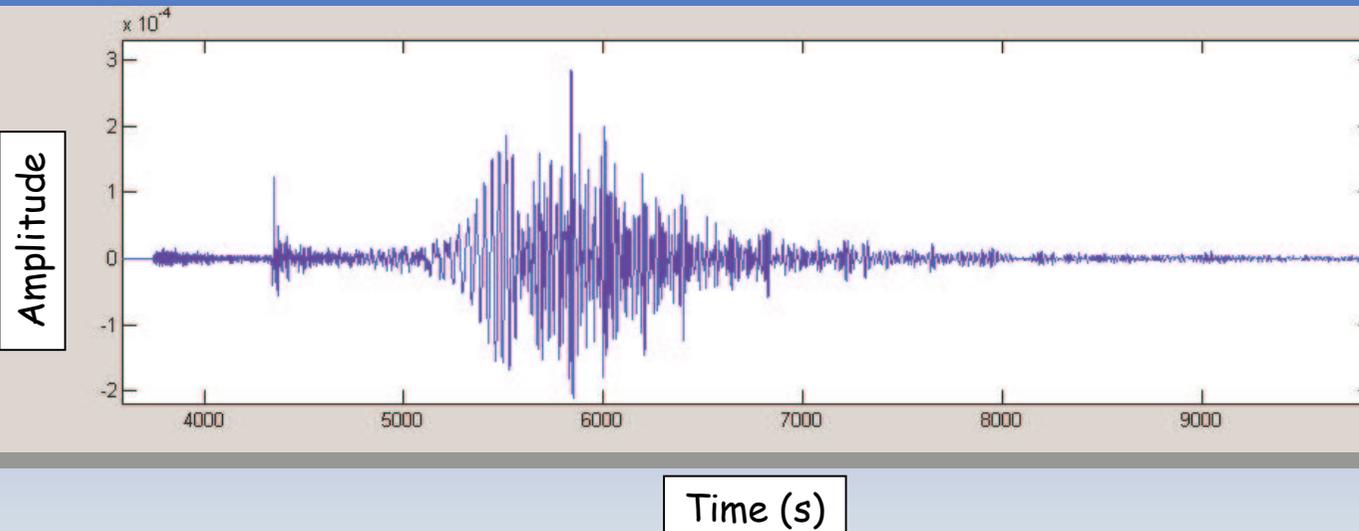
24 hour ground motion, do you see any signal?

Seismo-Weather



Running spectrum of the same data

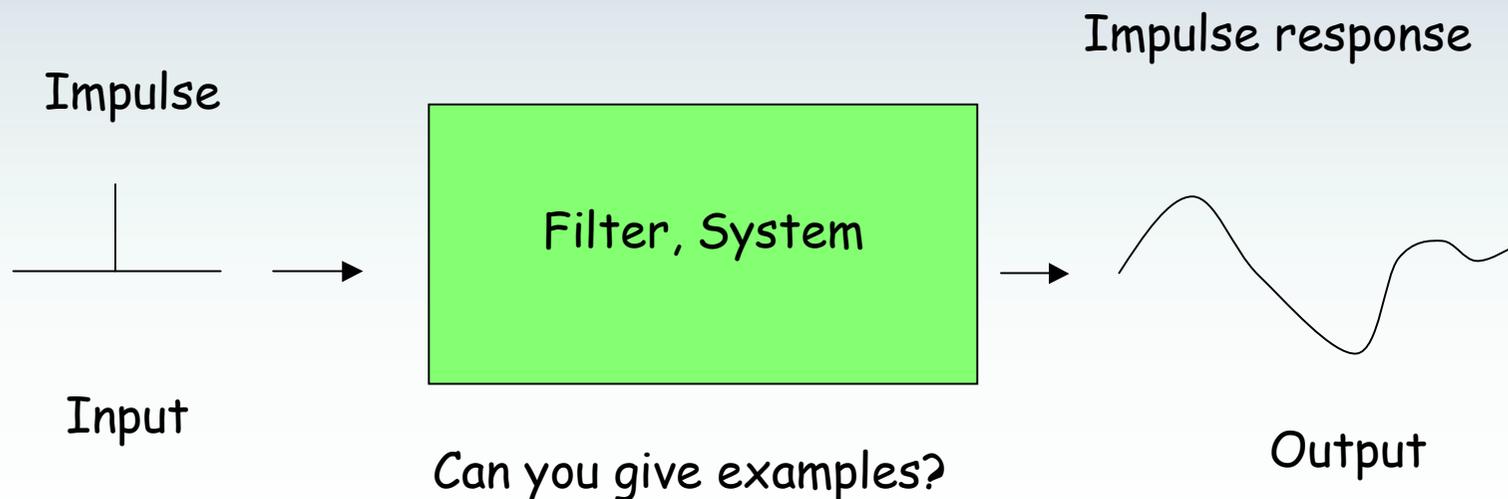
A seismogram



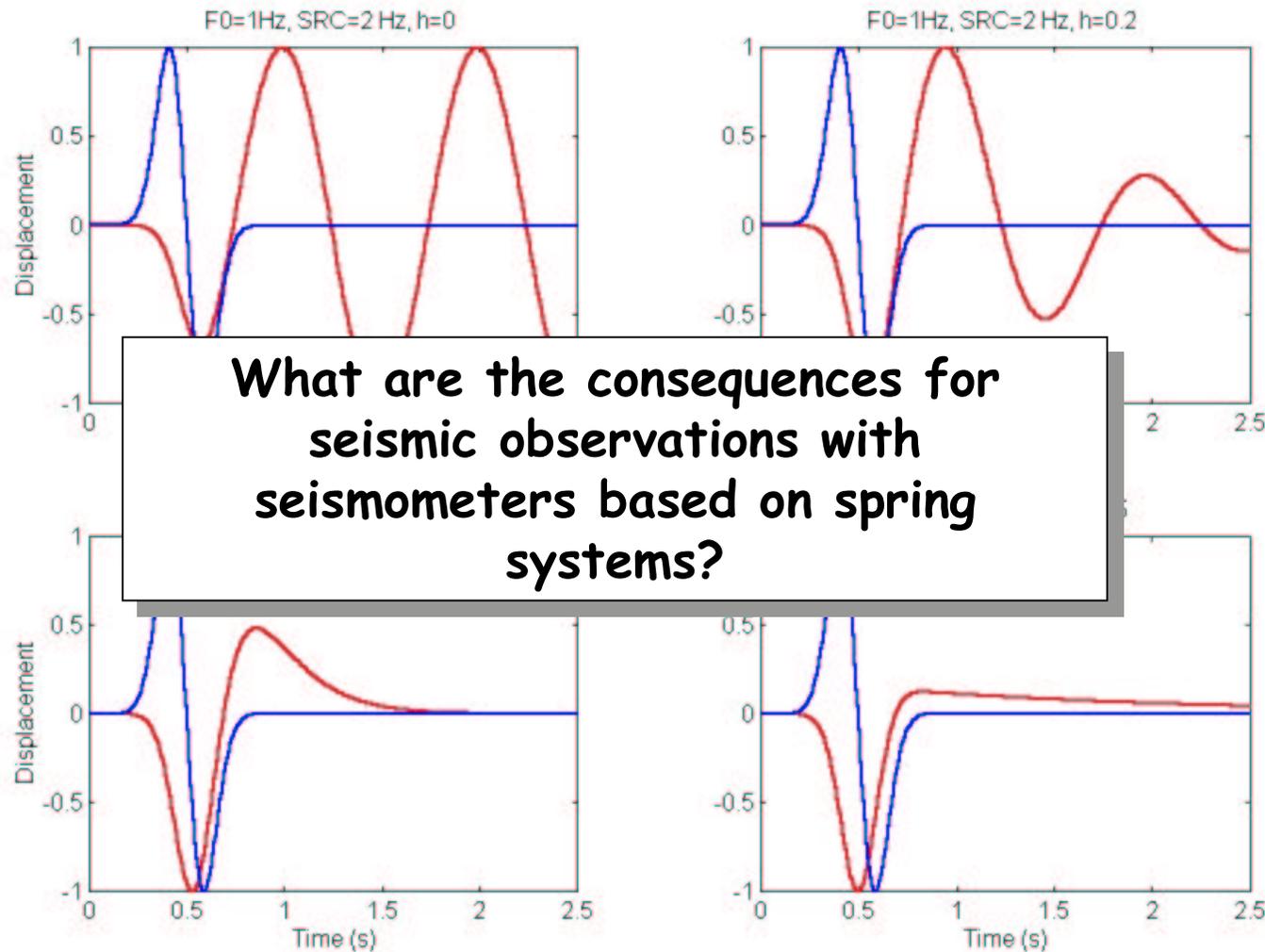
Waveform processing

How can (must) we treat our digitized observations to extract information on our physical signals? This question leads us to the concepts of (de-) convolution, (auto-,cross-) correlation and filtering.

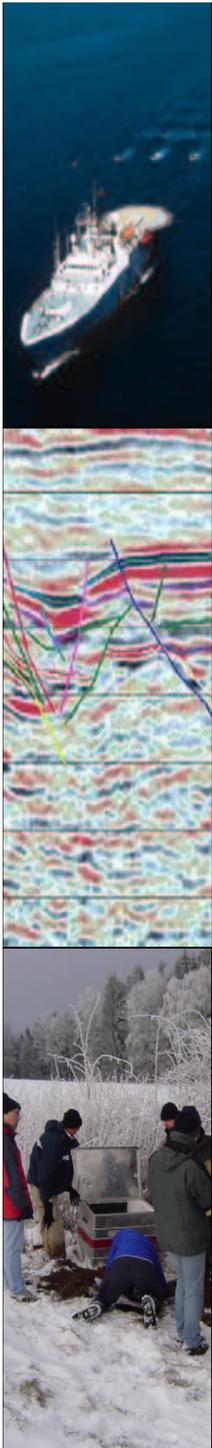
The central concept is the response of a system to an impulsive input; the **impulse response**



Impulse response of a seismometer



What are the consequences for seismic observations with seismometers based on spring systems?



Discrete Convolution

Convolution is the mathematical description of the change of waveform shape after passage through a filter (system).

There is a special mathematical symbol for convolution (*):

$$y(t) = g(t) * f(t)$$

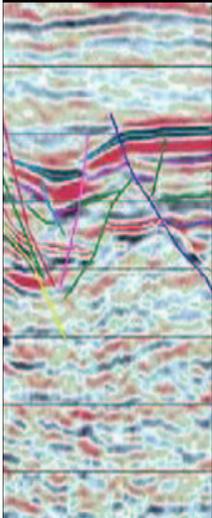
Here the impulse response function g is convolved with the input signal f . g is also named the „Green's function“

$$y_k = \sum_{i=0}^m g_i f_{k-i}$$

$$k = 0, 1, 2, \dots, m + n + 1$$

$$g_i \quad i = 0, 1, 2, \dots, m$$

$$f_j \quad j = 0, 1, 2, \dots, n$$

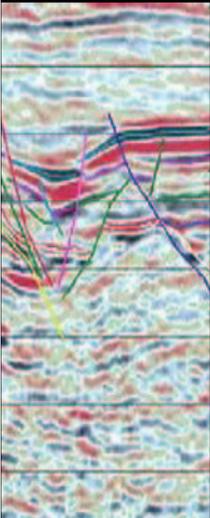


Convolution Example (Matlab)

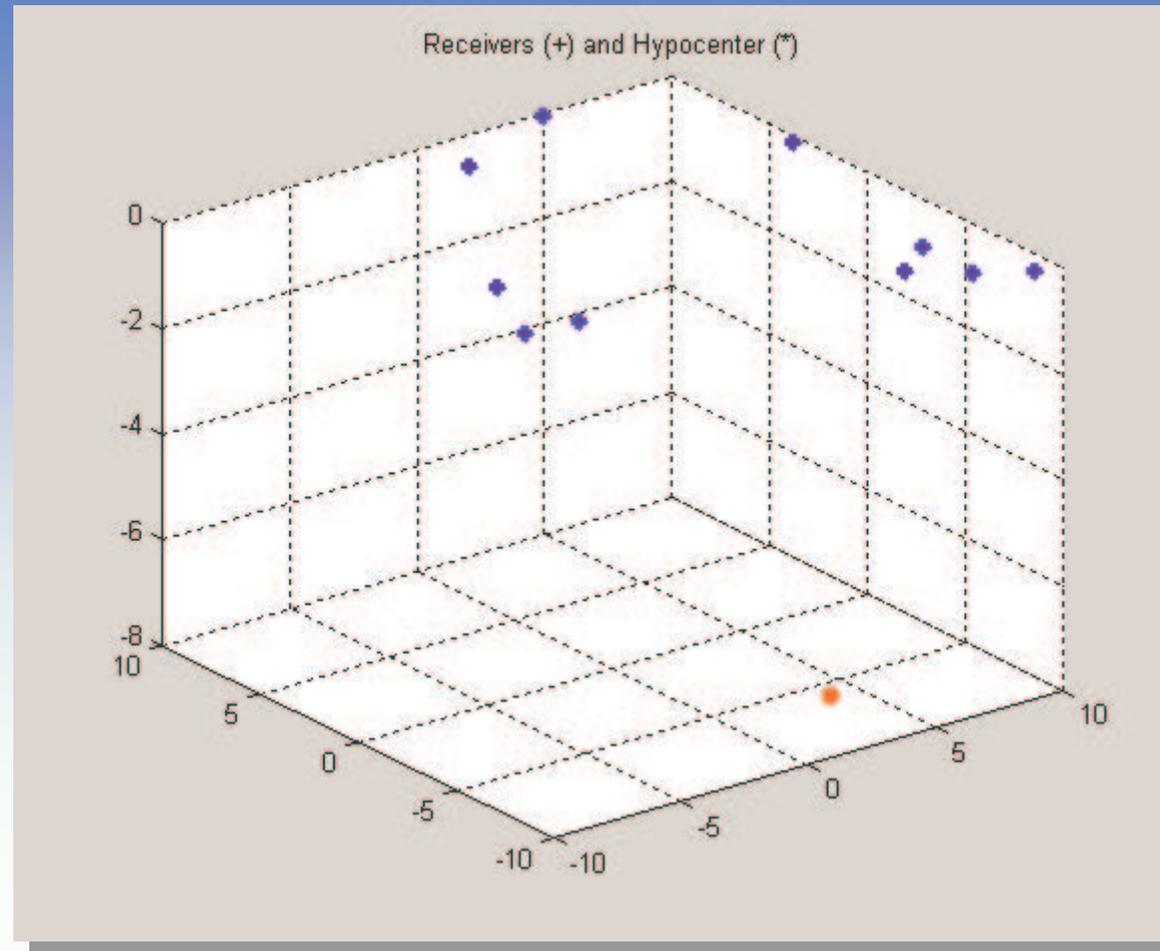
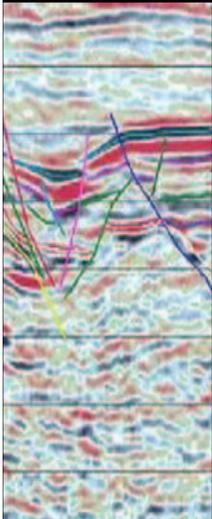
```
>> x
x =
    0    0    1    0

>> y
y =
    1    2    1

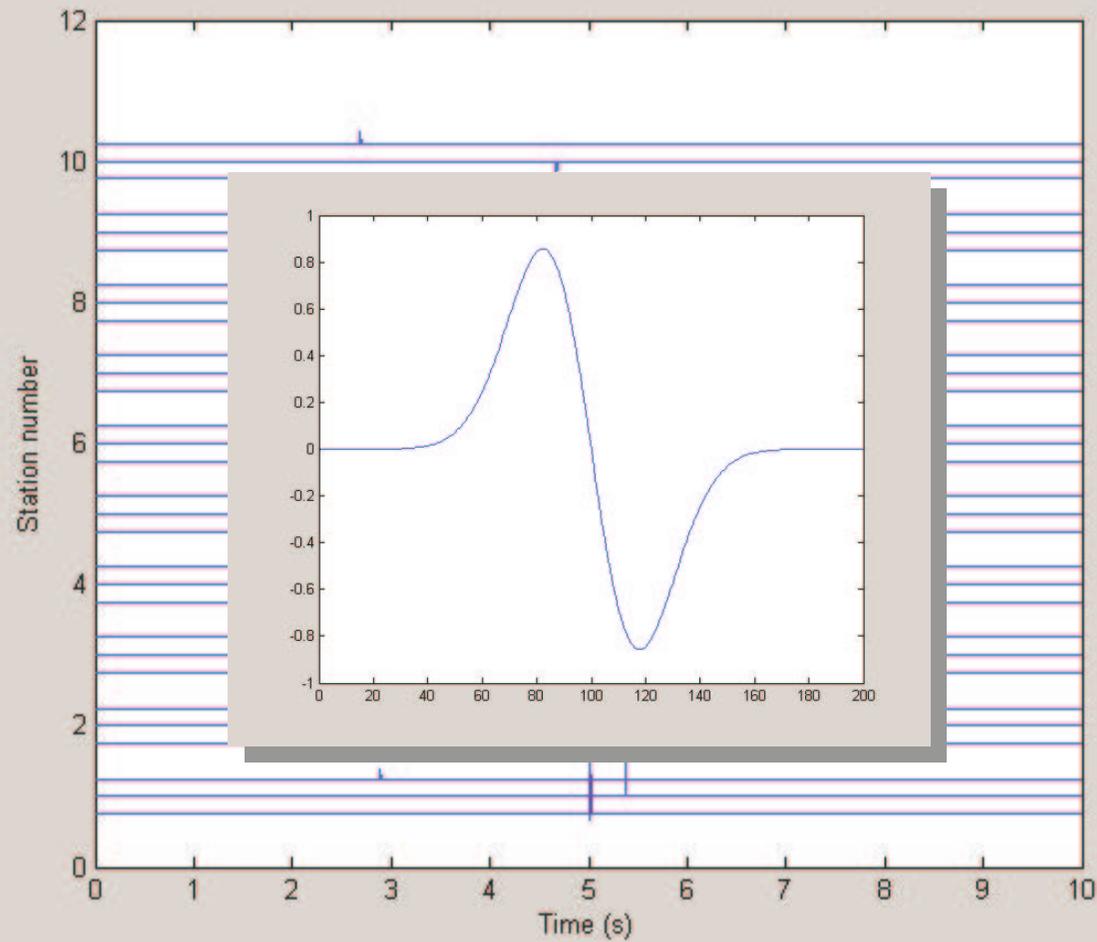
>> conv(x,y)
ans =
    0    0    1    2    1    0
```



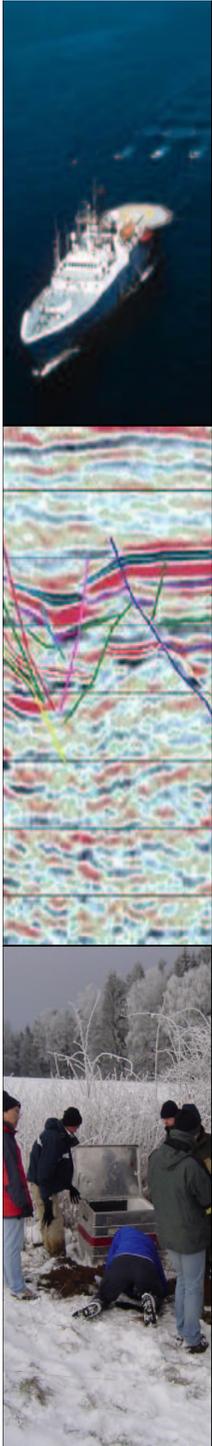
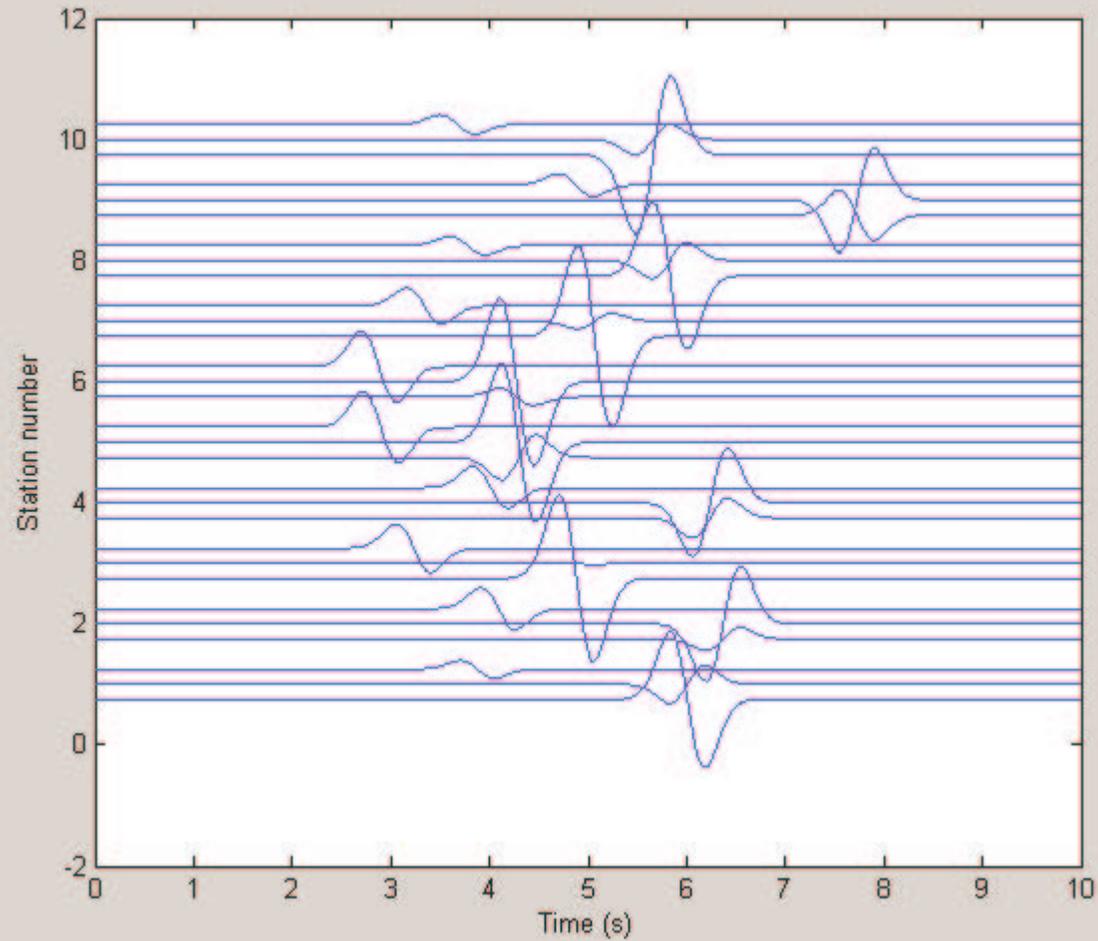
Convolutional model: *seismograms*



The seismic *impulse response*



The **filtered** response

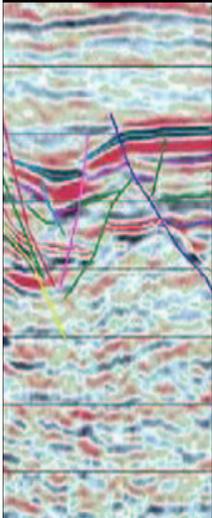
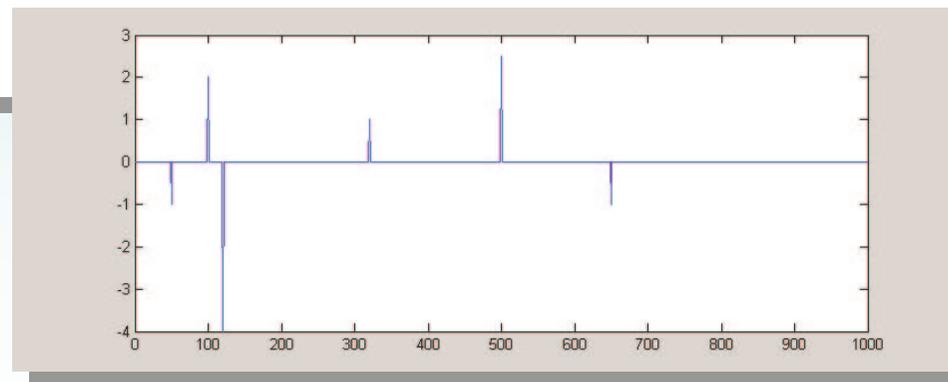
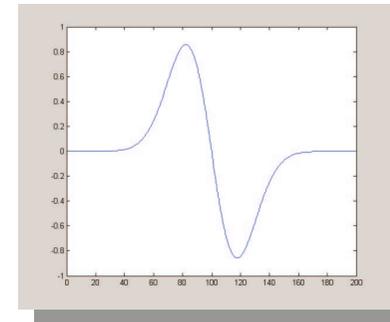


1D convolutional model of a seismic trace

The seismogram of a layered medium can also be calculated using a convolutional model ...

$$u(t) = s(t) * r(t) + n(t)$$

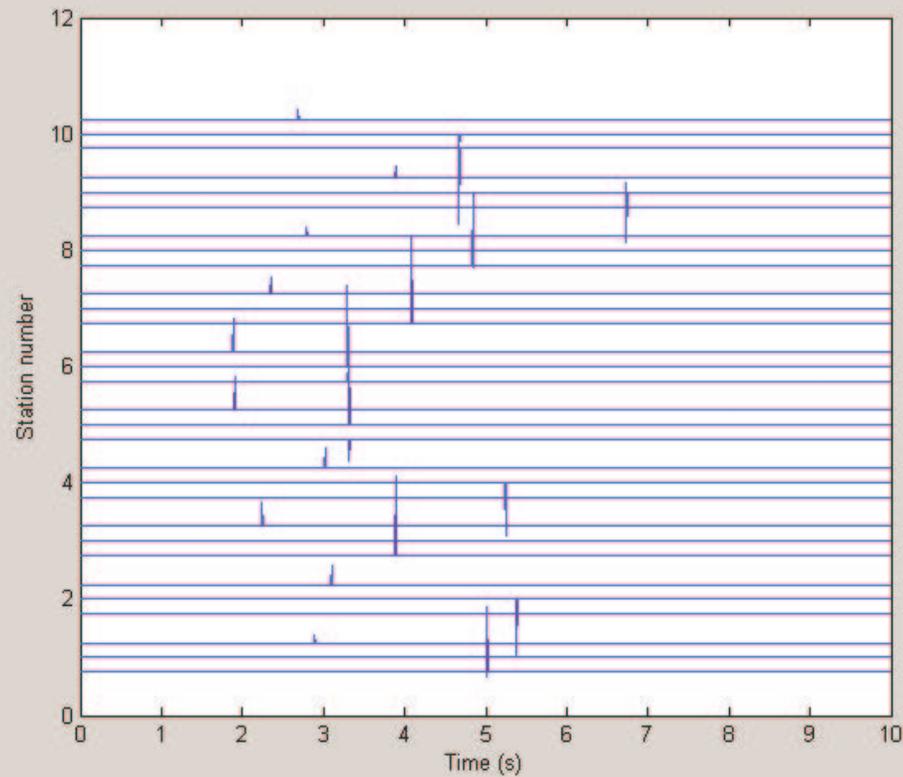
$u(t)$ seismogram
 $s(t)$ source wavelet
 $r(t)$ reflectivity



Deconvolution

Deconvolution is the inverse operation to **convolution**.

When is **deconvolution** useful?

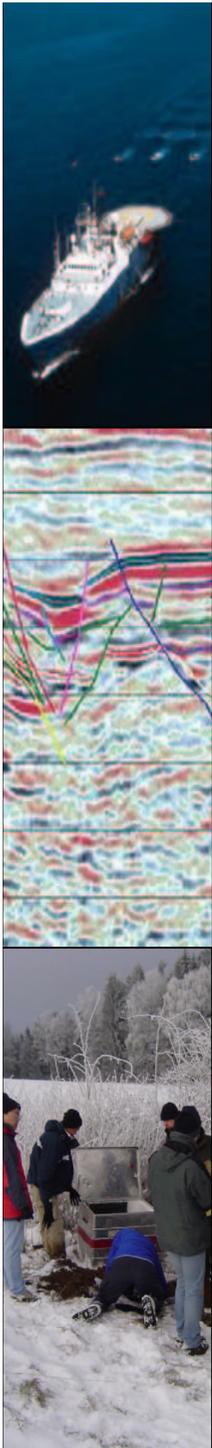


Correlation

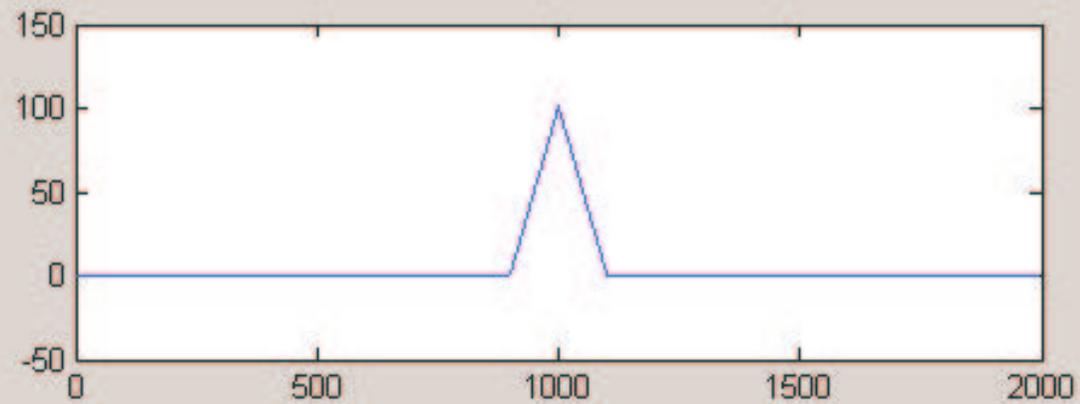
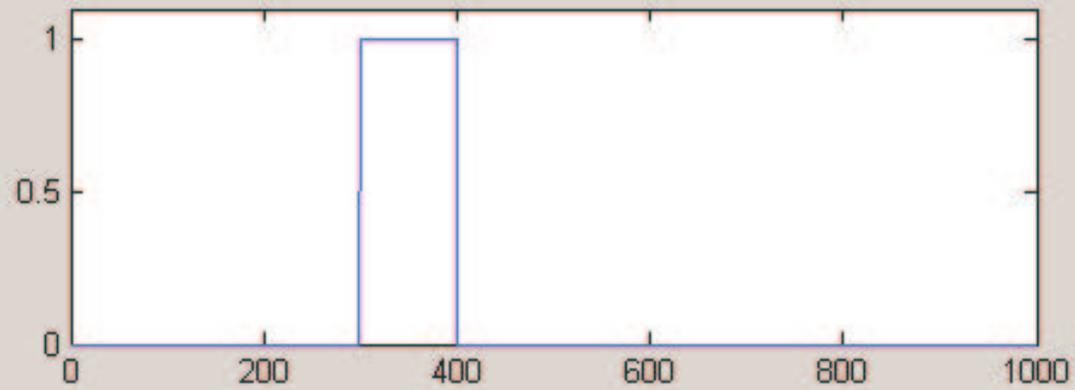
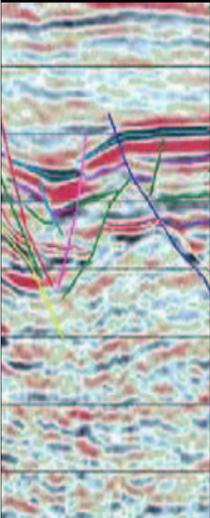
Correlation plays a central role in the study of time series. In general, correlation gives a **quantitative estimate of the degree of similarity between two functions**.

The correlation of functions g and f both with N samples is defined as:

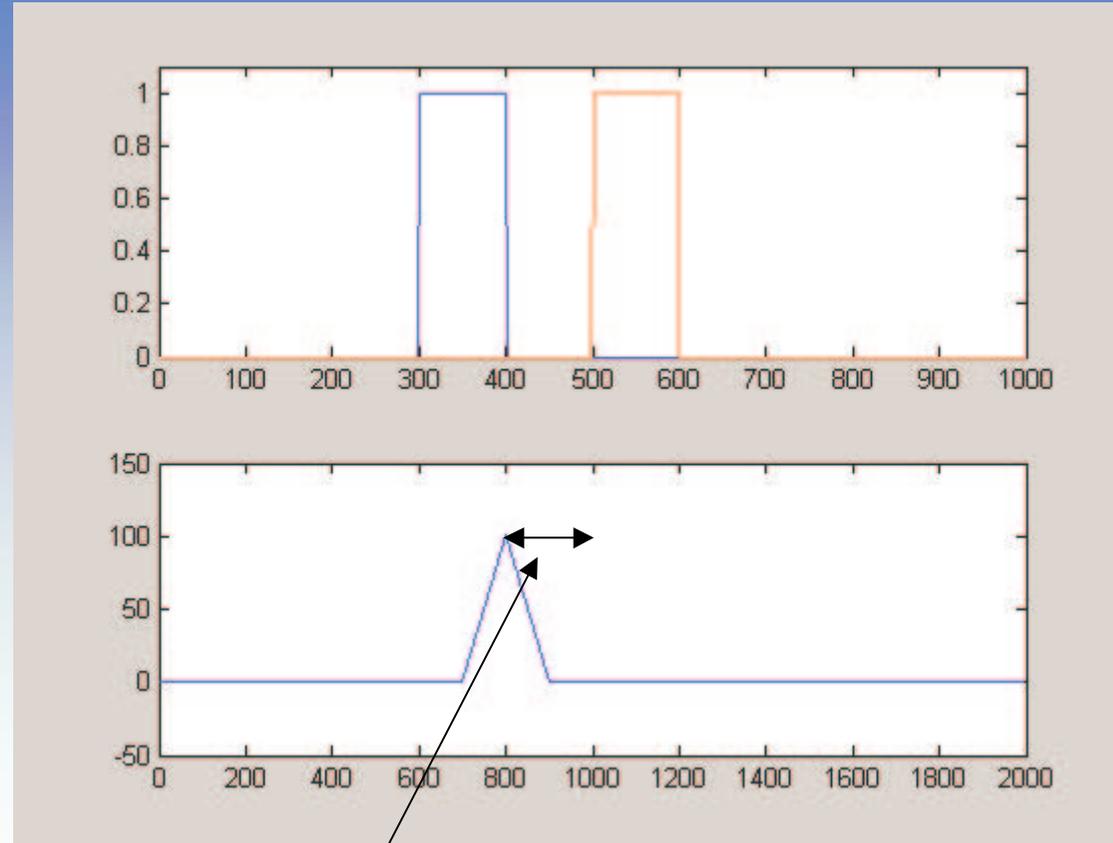
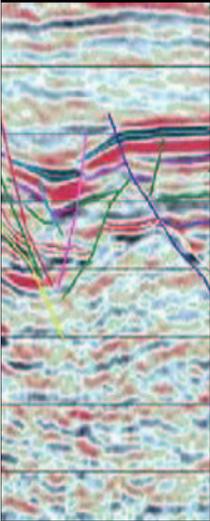
$$r_k = \frac{1}{N} \sum_{i=0}^{N-k-1} g_i f_{k+i}$$
$$k = 0, 1, 2, \dots, N - 1$$



Auto-correlation

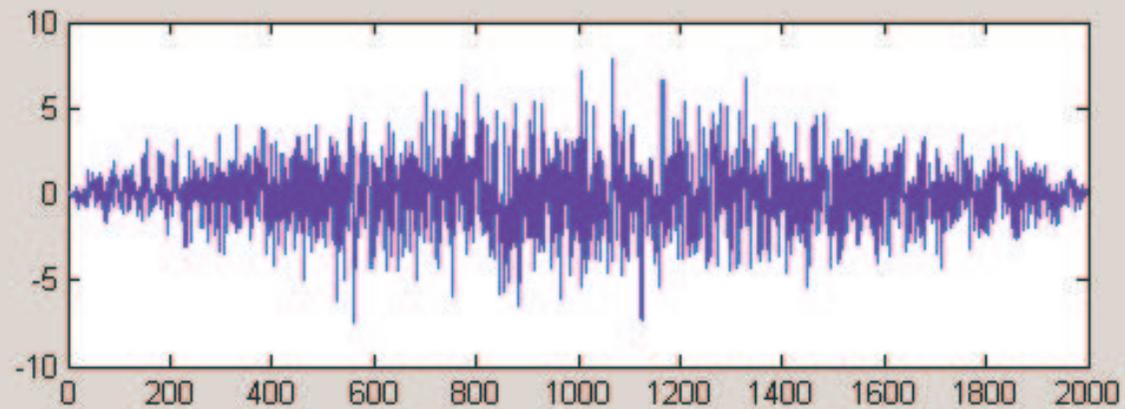
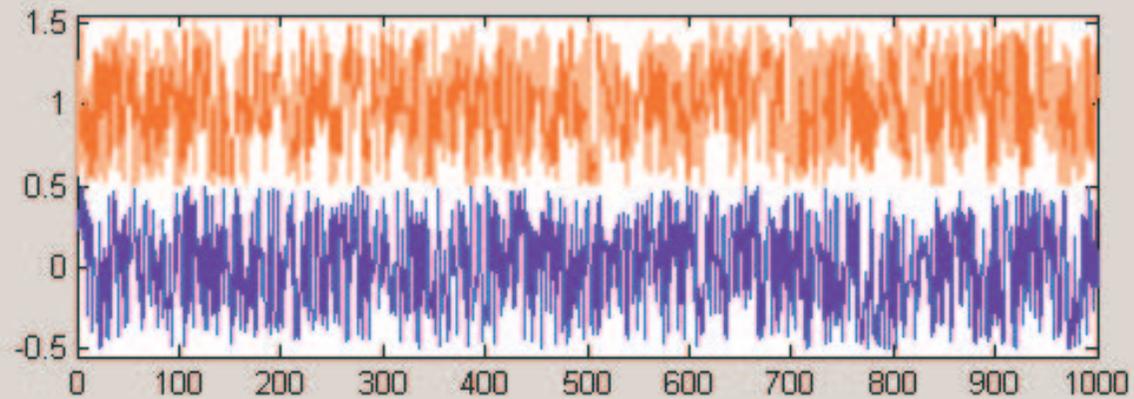
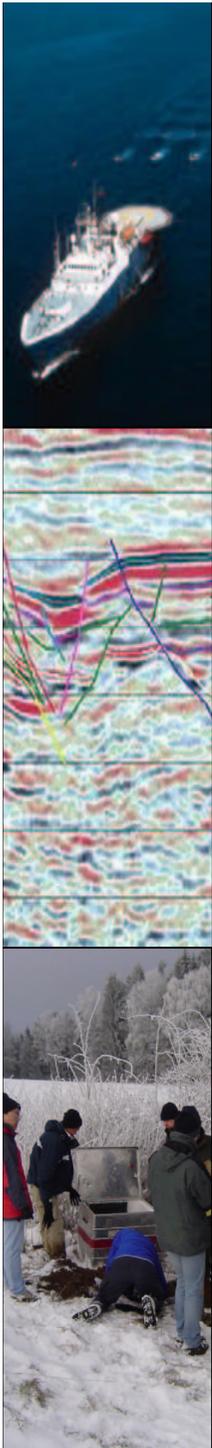


Cross-correlation

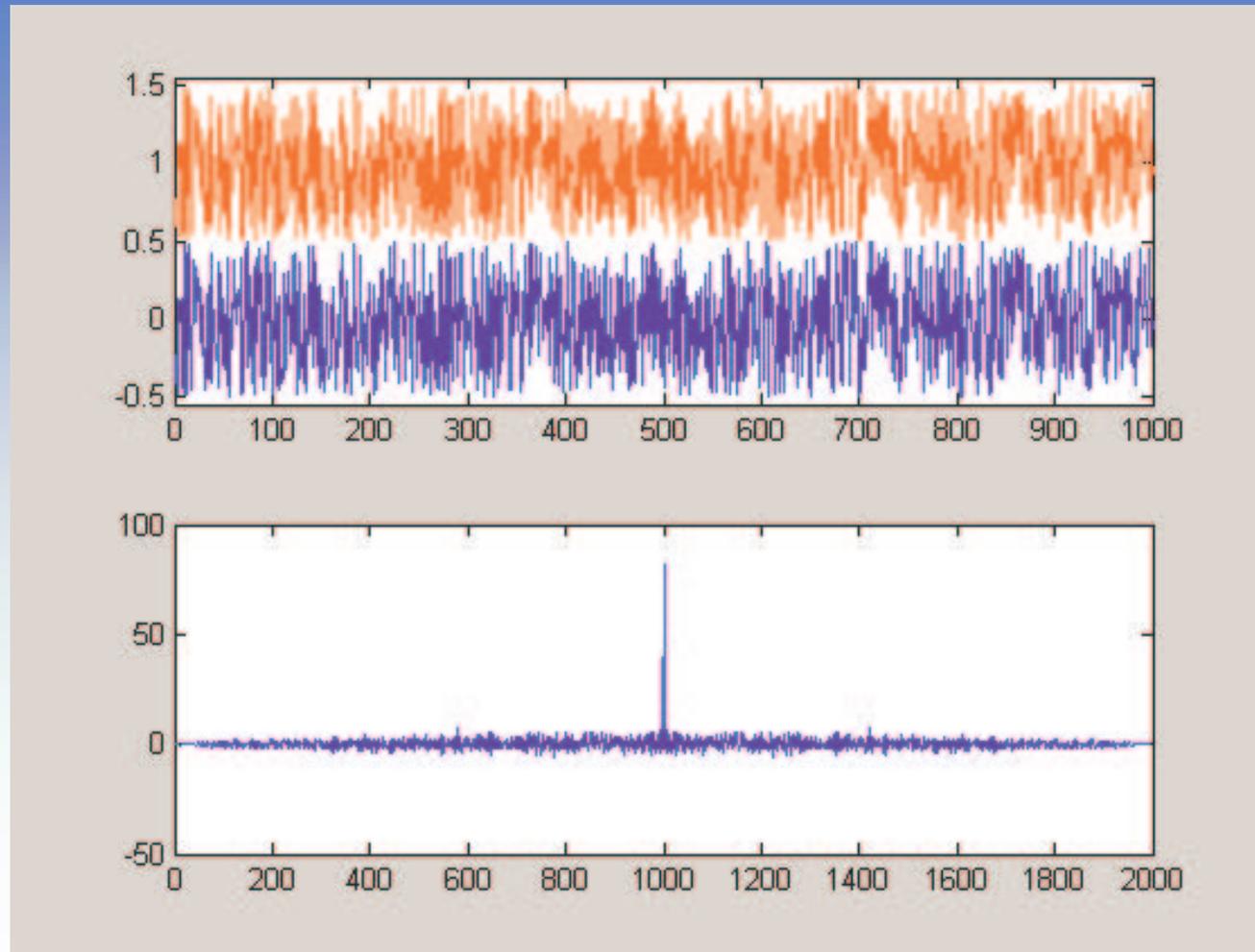
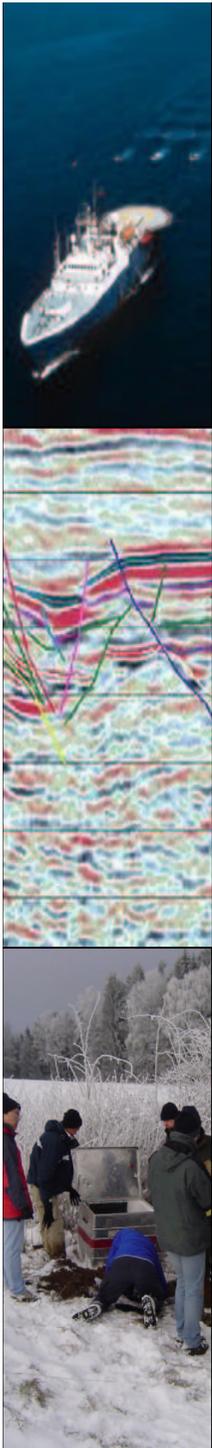


Lag between two functions

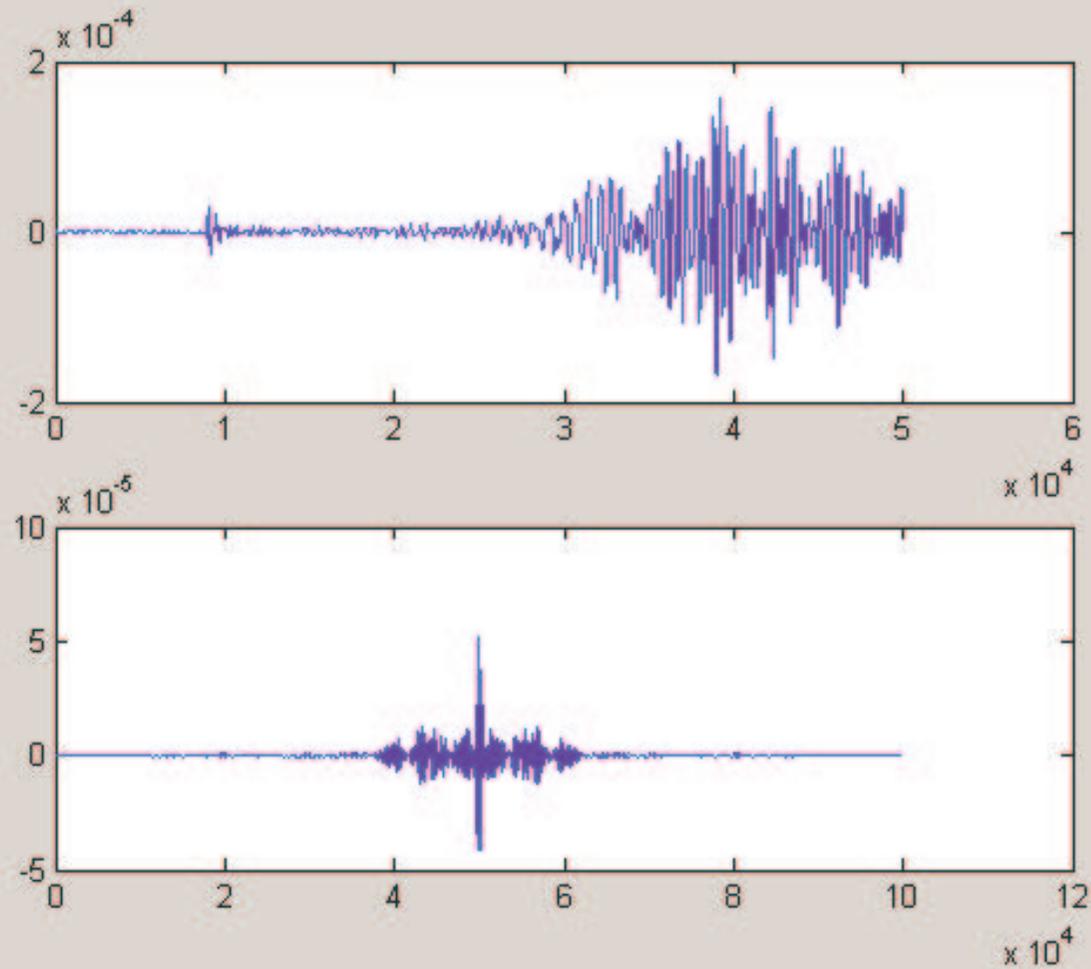
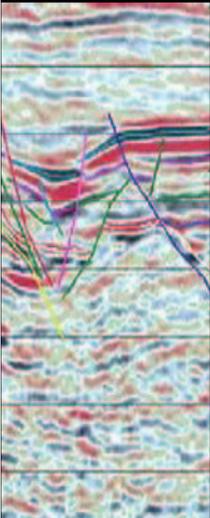
Cross-correlation Random functions



Auto-correlation Random functions

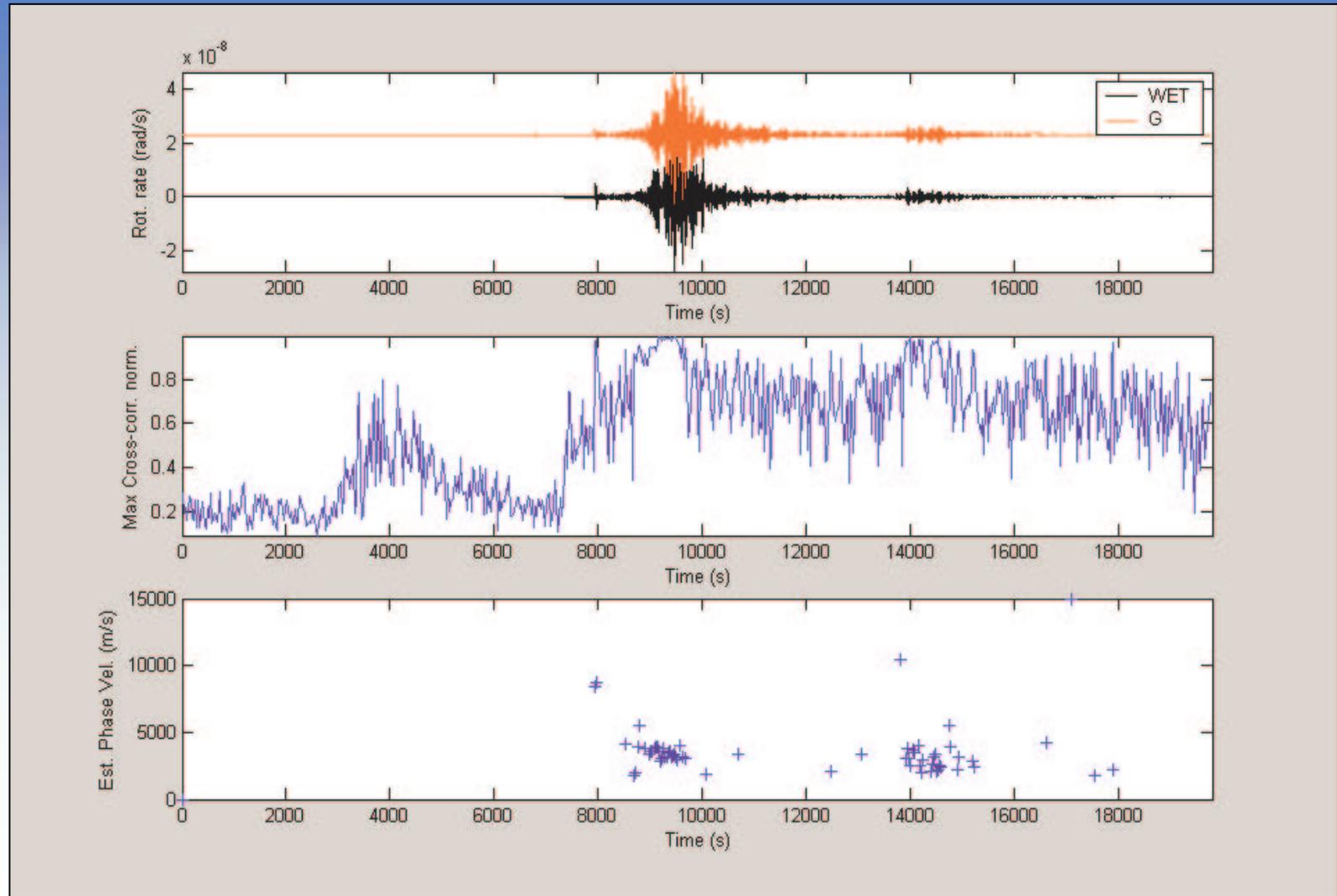
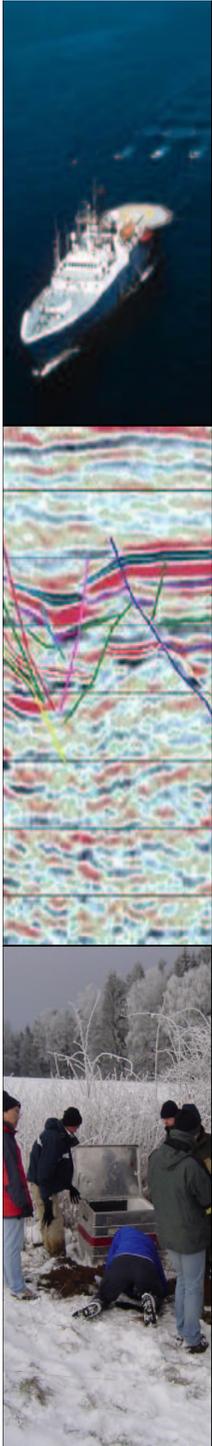


Auto-correlation Seismic signal



Cross-correlation

An example

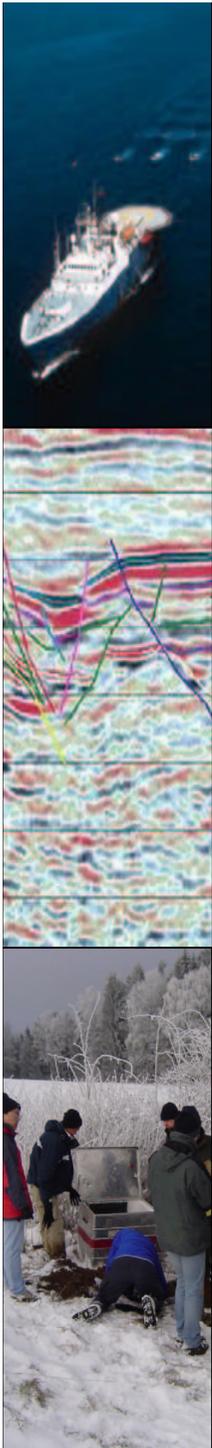


Digital Filtering

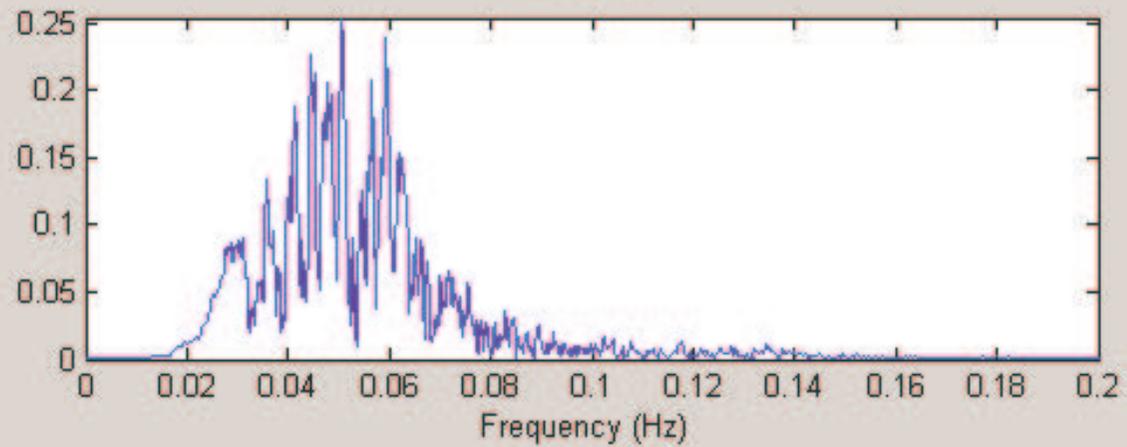
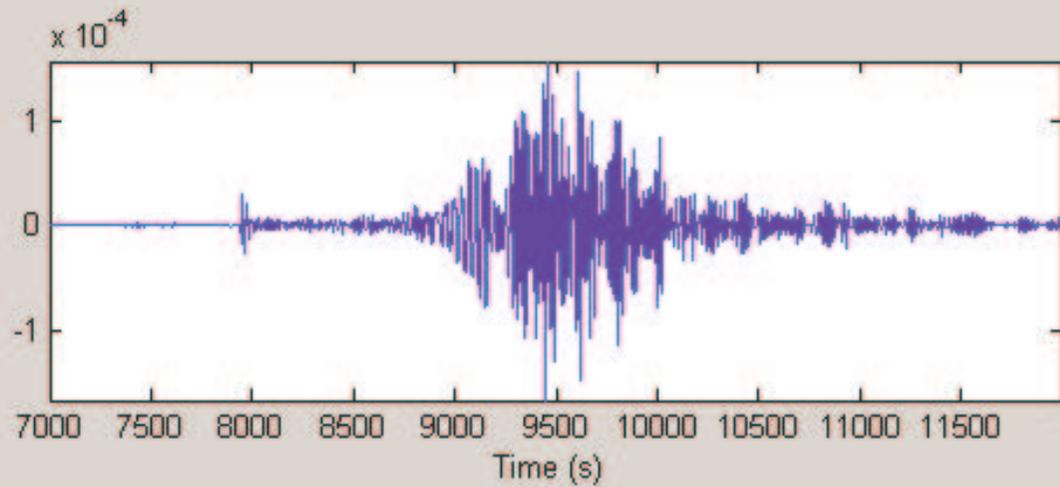
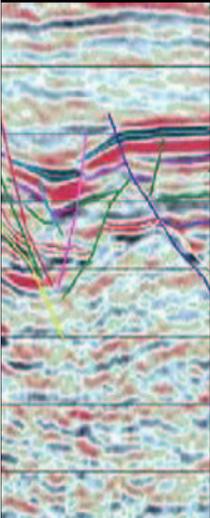
Often a recorded signal contains a lot of information that we are not interested in (noise). To get rid of this noise we can apply a **filter in the frequency domain**.

The most important filters are:

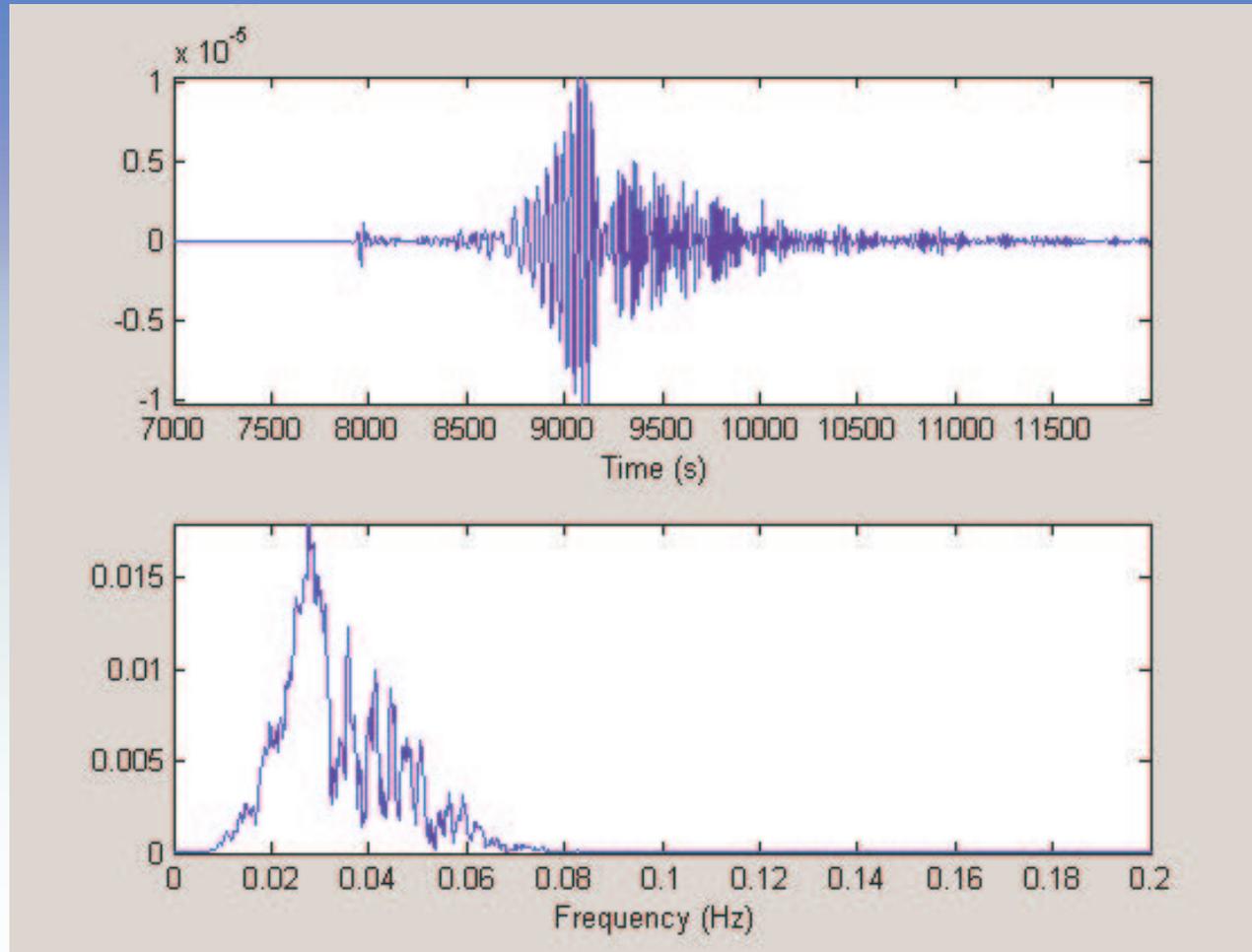
- **High pass:** cuts out low frequencies
- **Low pass:** cuts out high frequencies
- **Band pass:** cuts out both high and low frequencies and leaves a band of frequencies
- **Band reject:** cuts out certain frequency band and leaves all other frequencies



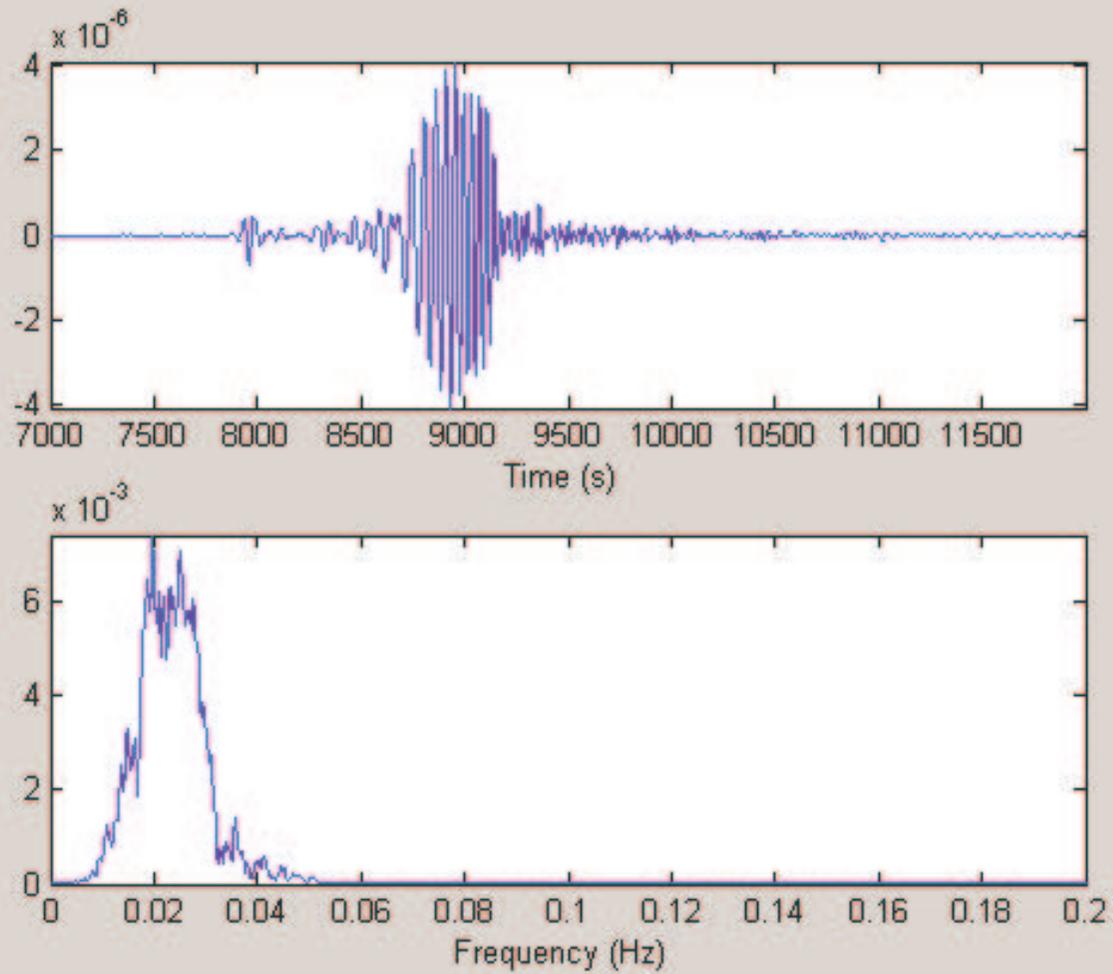
Digital Filtering



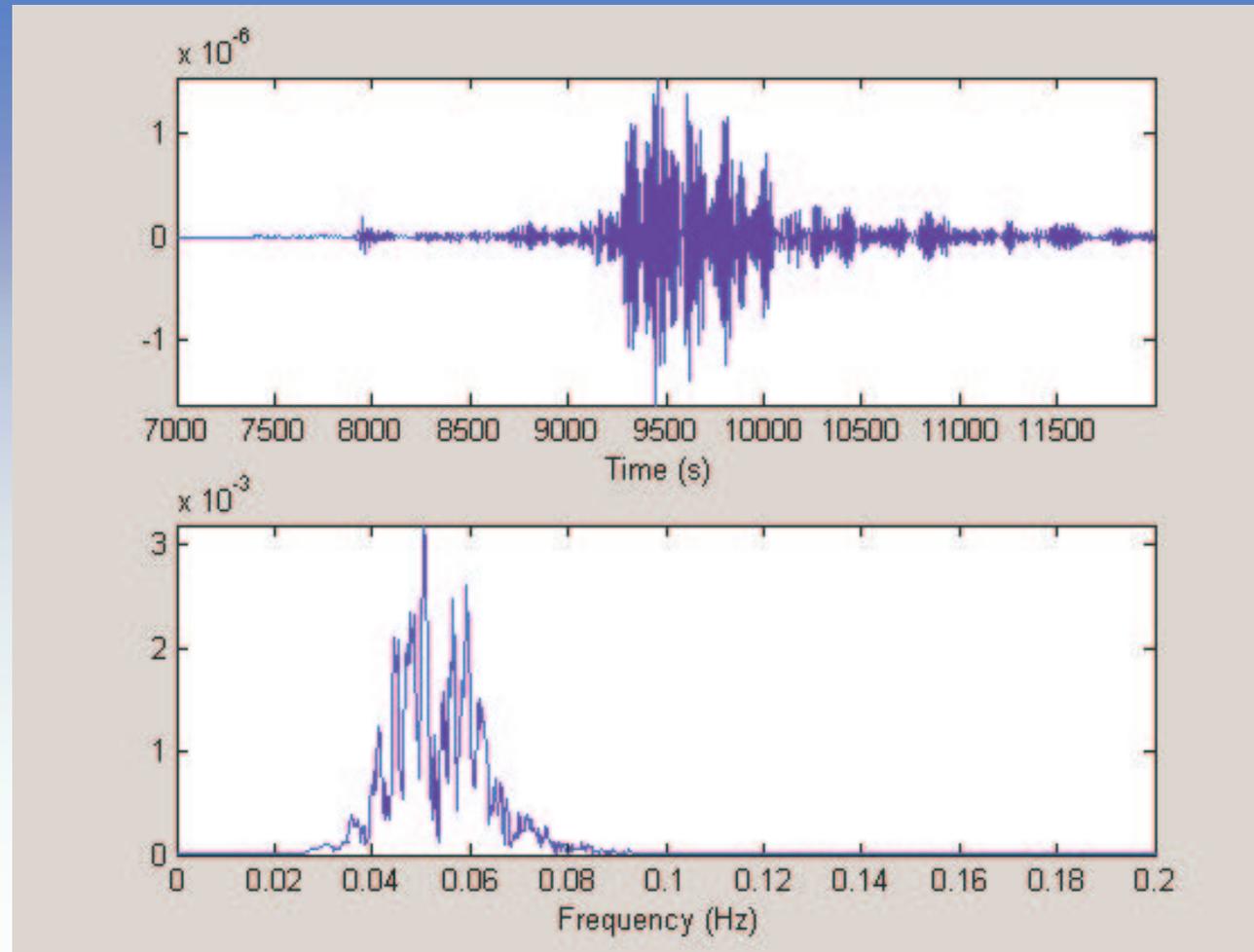
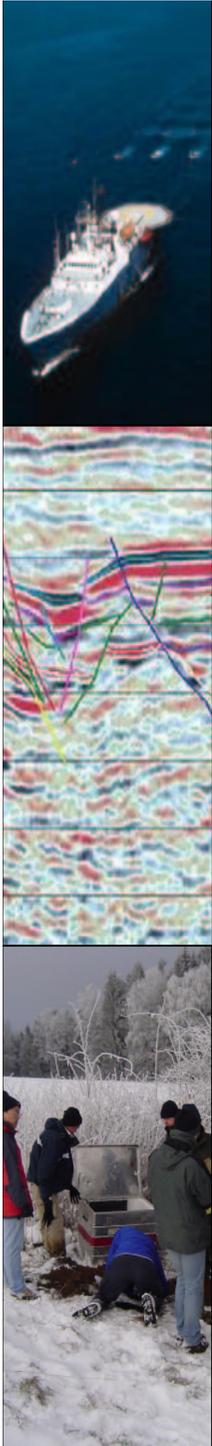
Low-pass filtering



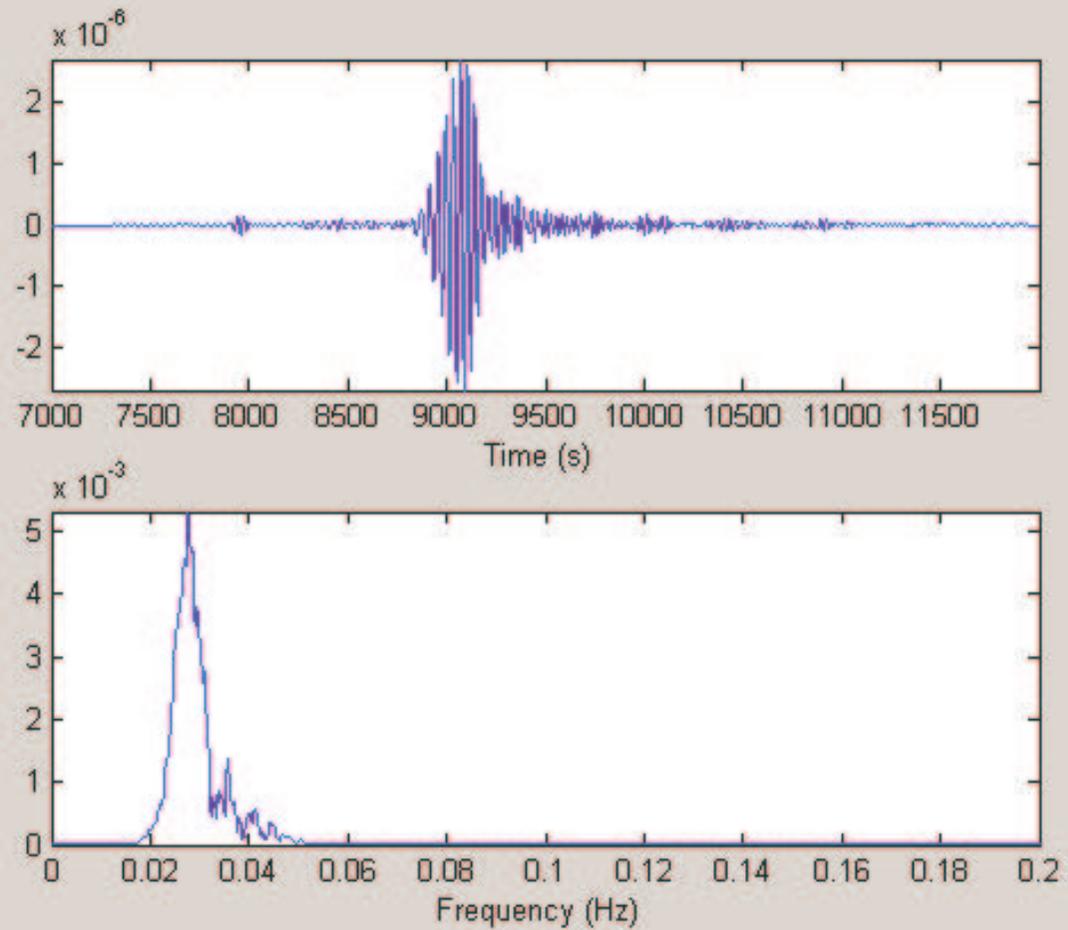
Lowpass filtering



High-pass filter



Band-pass filter



Summary

Today almost all data analysis involves the use of spectral methods and filtering.

The concepts are:

Convolution -> to obtain the response of a system to a particular input

Correlation -> to compare signal for their similarity

Fourier transformation - spectra - filtering -> to cut out certain frequencies and potentially highlight the signals of interest

