

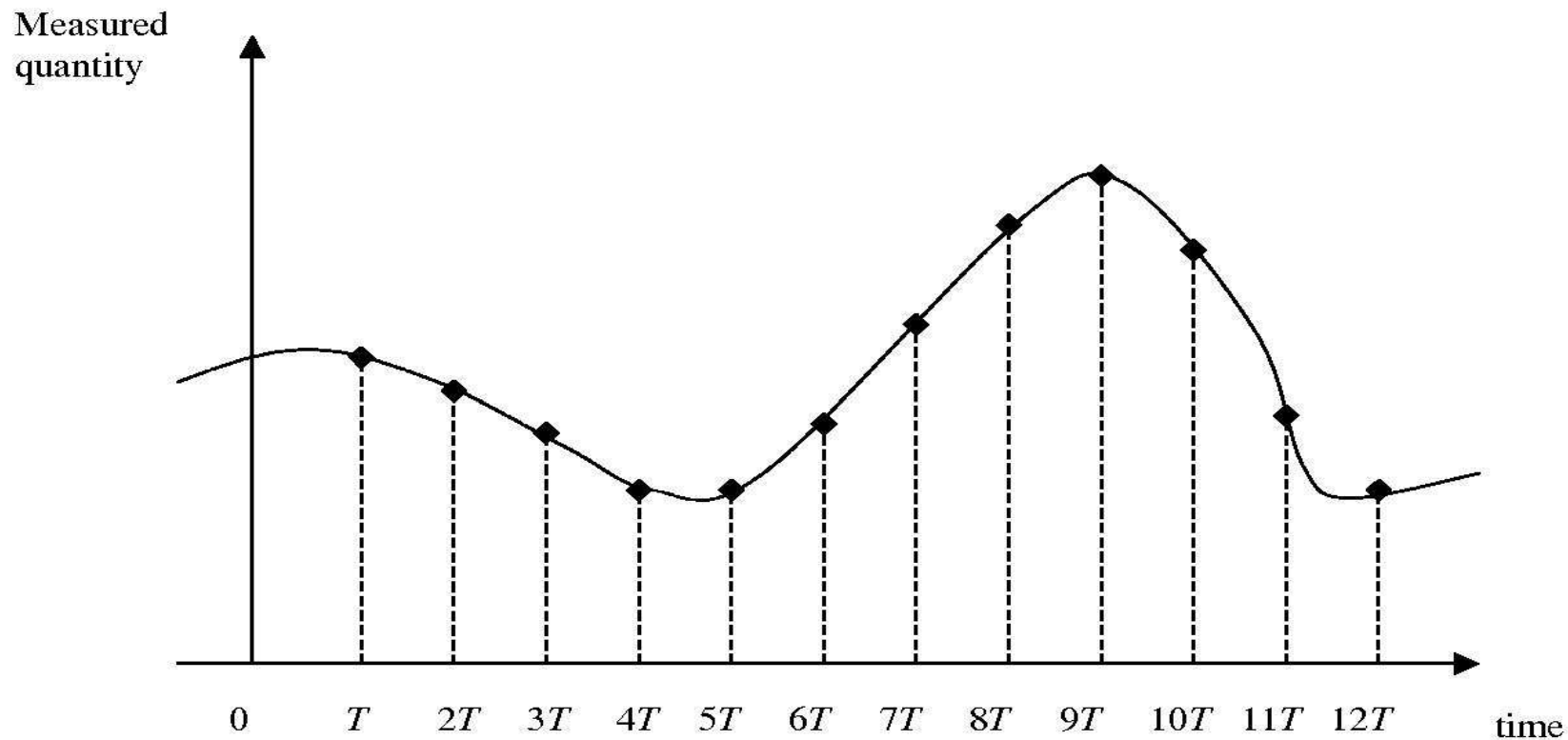
# First steps in digital signal processing

John Coleman

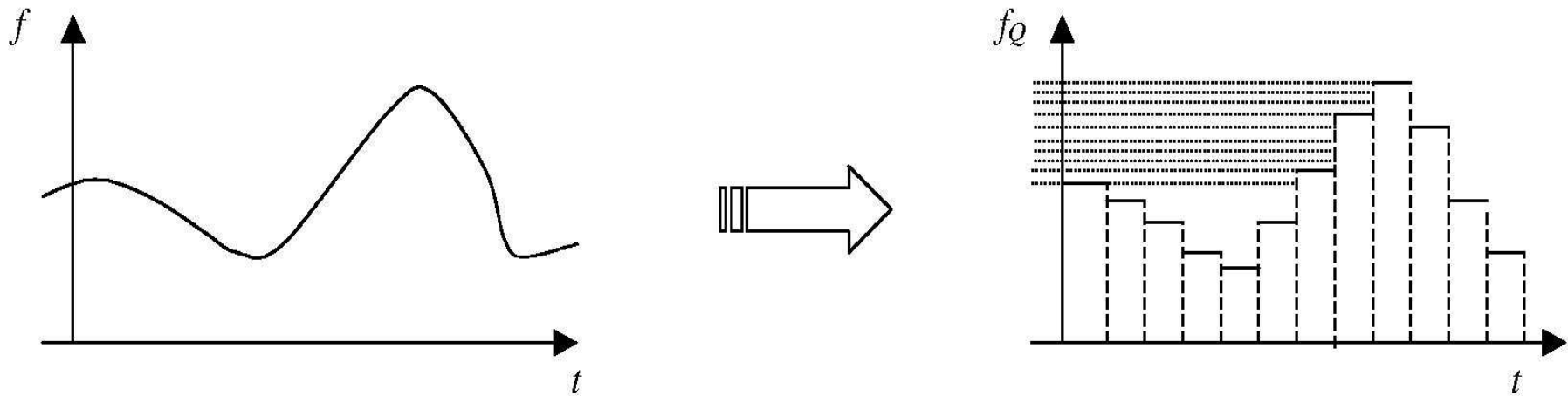
Phonetics Laboratory  
University of Oxford



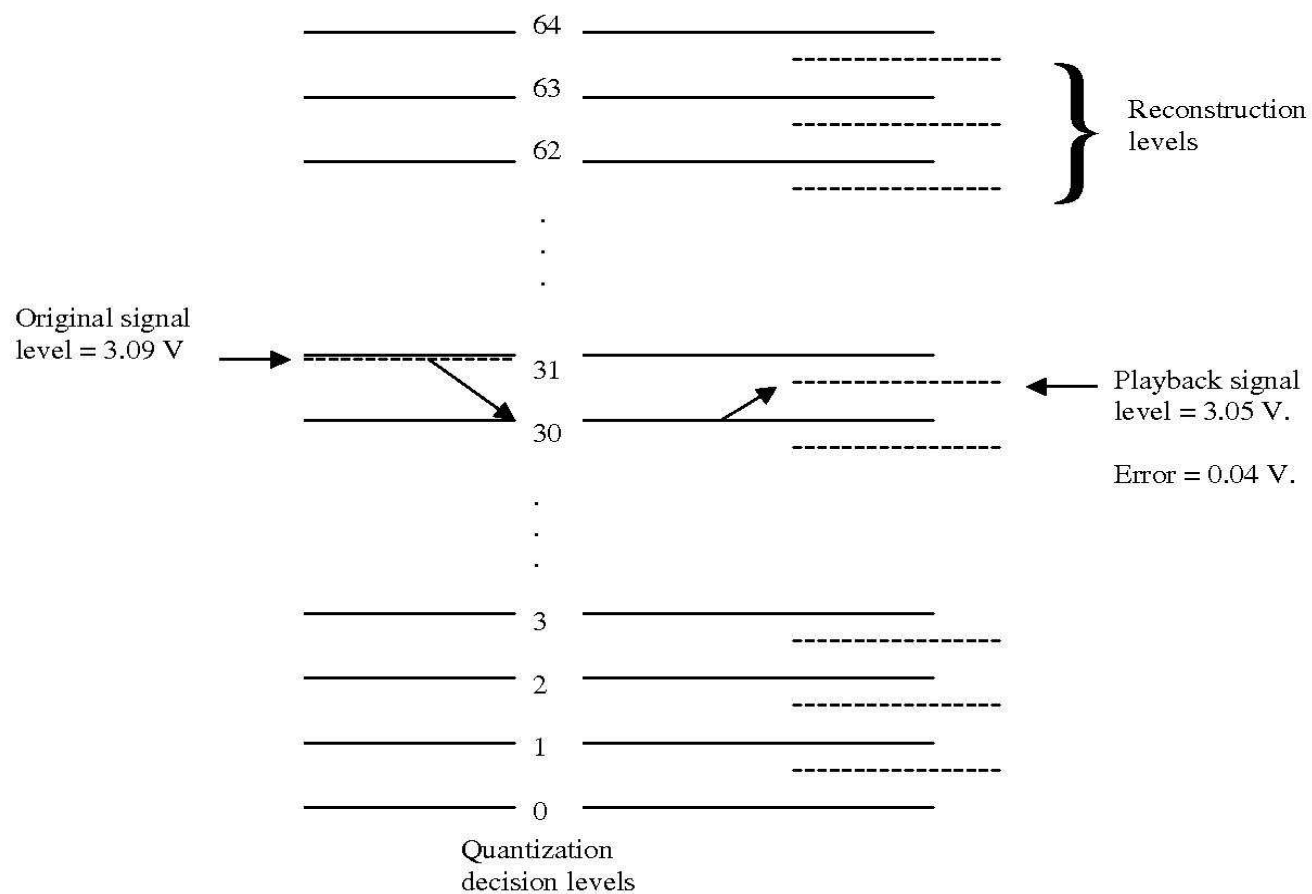
# Analogue-to-digital conversion 1: Sampling



# Analogue-to-digital conversion 2: Quantization



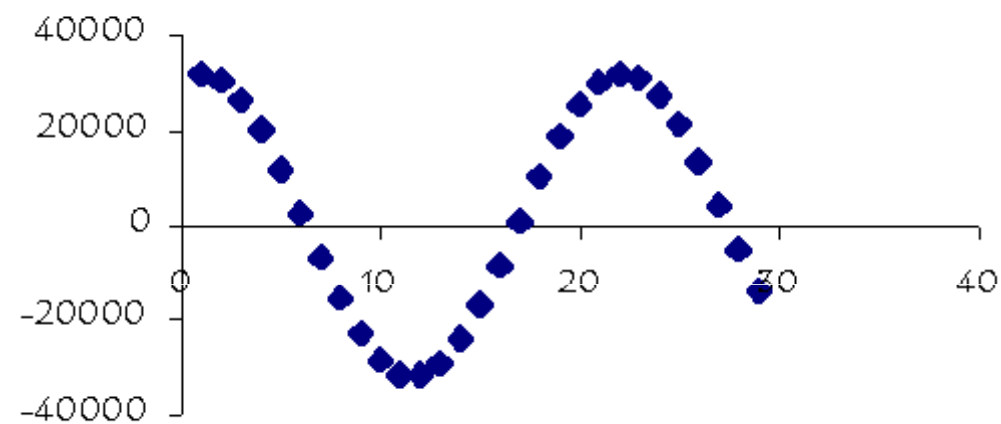
# D-to-A, quantization error



# D-to-A, quantization error

- To reduce the quantization error, use more levels (dynamic range):
  - 8 bits:  $2^8 = 256$  levels
  - 12 bits:  $2^{12} = 4096$  levels
  - 16 bits:  $2^{16} = 65536$  levels

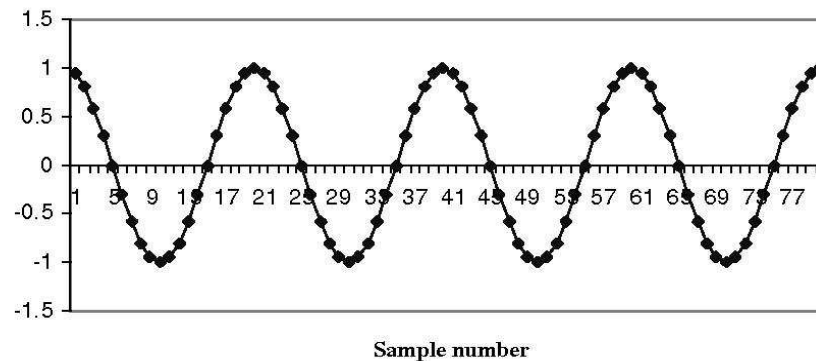
# Pulse Code Modulation (Alec Reeves 1937)



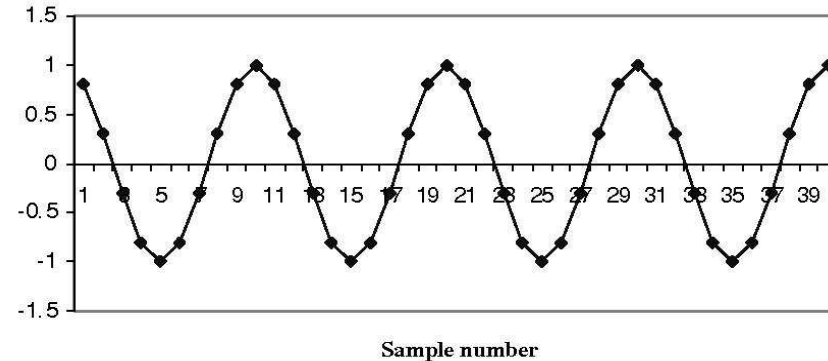
# Sampling theorem; Nyquist frequency

- Sampling rate must be at least twice the highest frequency you want to capture

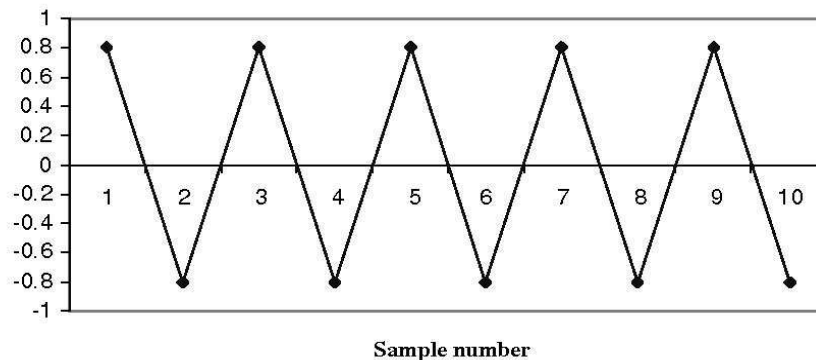
(a) 20 samples per cycle



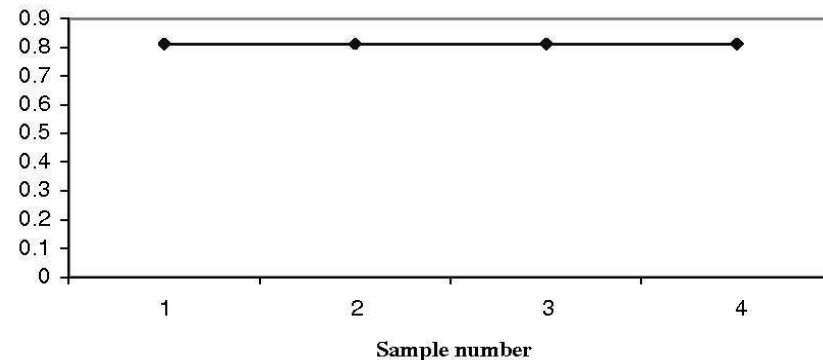
(b) 10 samples per cycle



(c) 2 samples per cycle



(d) 1 sample per cycle



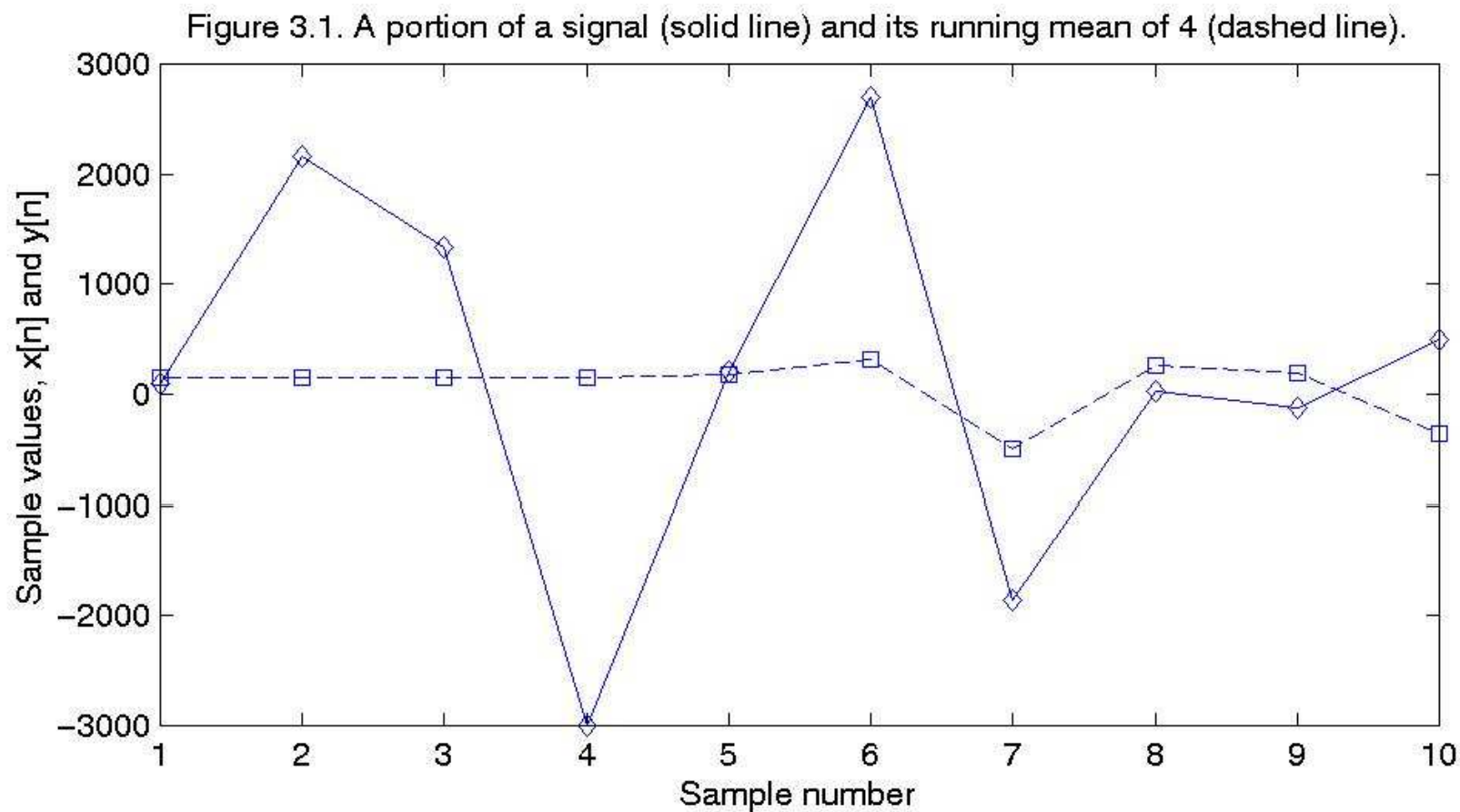
# Operations on sequences of numbers

- Let's call the sample number  $i$  and the  $i$ 'th sample  $x[i]$
- **Sum** or **integral**,  $\Sigma x[i]$ .
- If  $x[i]$  has positive and negative values, take  $|x[i]|$  the absolute (i.e. unsigned) value of  $x[i]$ .
- Or, first calculate the square of  $x[i]$ ,  $x[i]^2$ , and then take the square root,  $\sqrt{x[i]^2}$ .  $\Sigma \sqrt{x[i]^2}$  is a measure of the overall energy of a signal.
- $x[i]^2$  gets bigger and bigger as  $x[i]$  gets longer.
- The *average* amplitude of a signal, calculated over  $n$  samples:  $\sqrt{(\Sigma x[i]^2)/n}$ . This is called the *root mean square* or RMS amplitude.



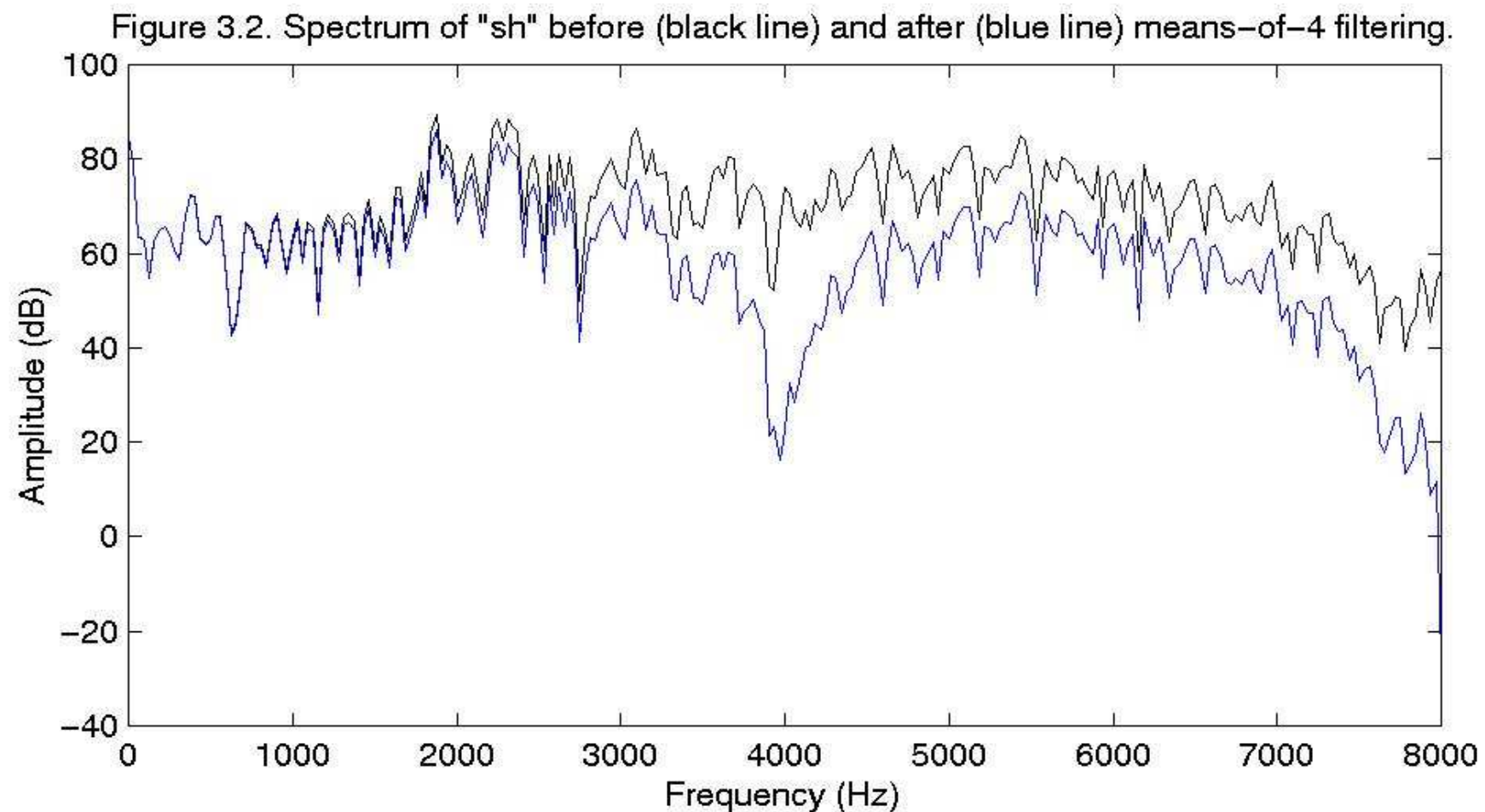
# Local (moving) average

e.g.,  $y[n] = \frac{1}{4} x[n] + \frac{1}{4} x[n-1] + \frac{1}{4} x[n-2] + \frac{1}{4} x[n-3]$



# Local (moving) average

## Effect: low-pass filtering



# Time-domain filtering

4 samples is very short, so its effects are very local – high frequency components. To smooth over a larger slice of the signal, we can do two things:

- a) increase the number of samples in  $x[n] \dots x[n-m]$
- b) make  $y[n]$  depend in part on its own previous value,  $y[n-1]$ , or several previous values.

# Time-domain filtering

- A filter of the second kind has the general form:

$$y[n] = b_0 x[n] + b_1 x[n-1] + b_2 x[n-1] + \dots + b_k x[n-k] \\ - a_1 y[n-1] + \dots + a_j y[n-j]$$

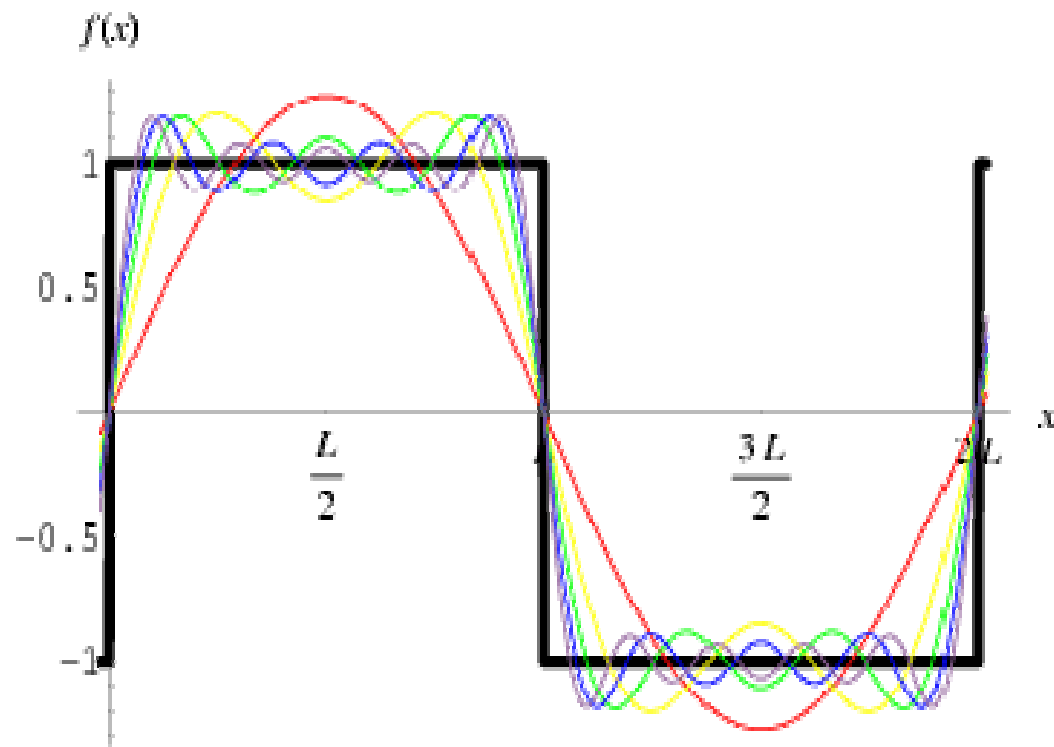
- By varying the a's, b's, and the number and spacing of previous x and y samples, a variety of filters with various kinds of frequency-selecting behaviours can be constructed.

# Linear Prediction (preview)

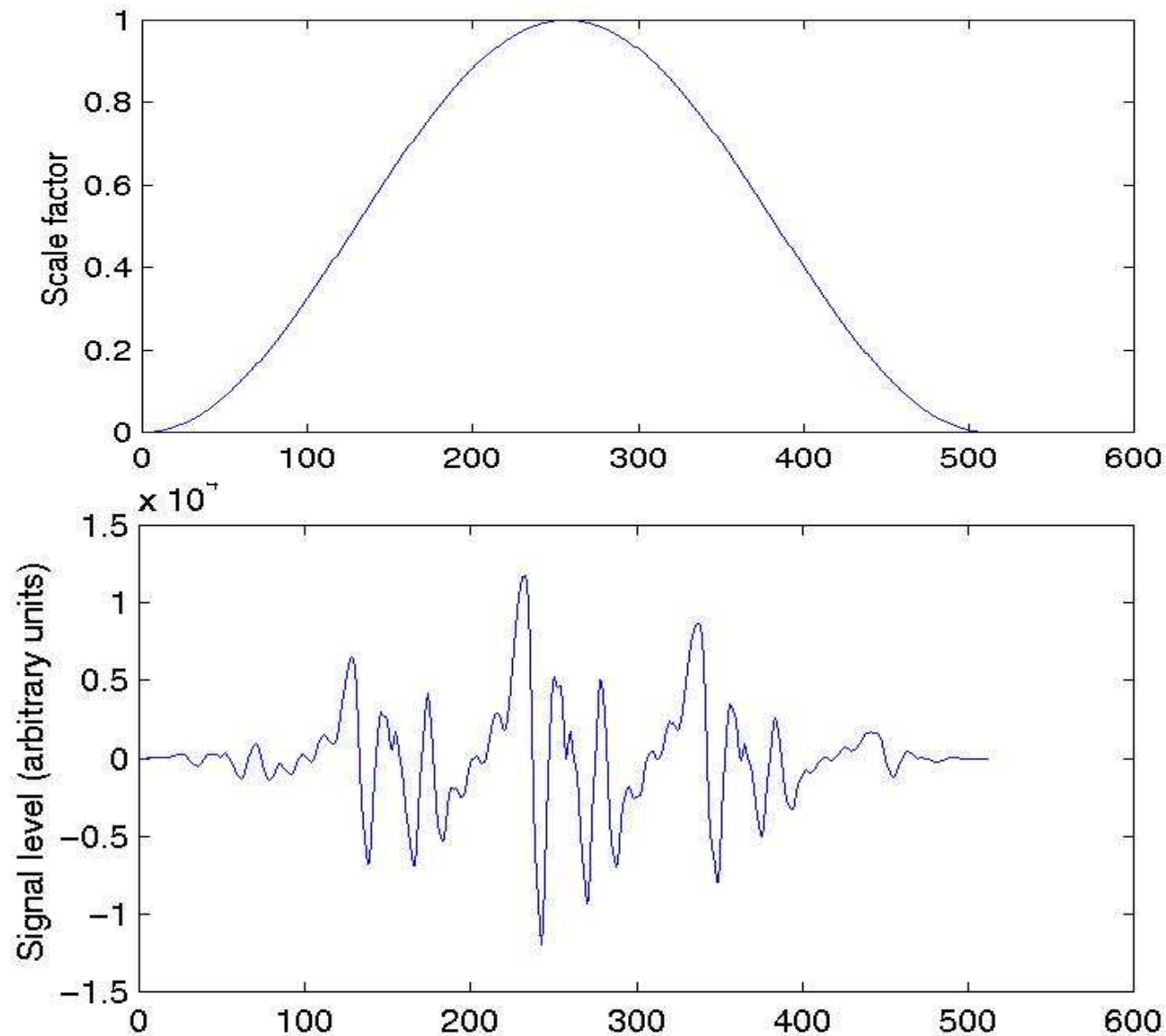
- $$y[n] = b_0 x[n] + b_1 x[n-1] + b_2 x[n-1] + \dots + b_k x[n-k] - a_1 y[n-1] + \dots + a_j y[n-j]$$
- We can estimate the magnitude of the current sample as a linear combination of the previous  $p$  samples (typically 12 to 18 samples):
- $$x[t] = -a_1 x[t-1] - a_2 x[t-2] - a_3 x[t-3] \dots - a_p x[t-p] + e[t]$$

# Fourier (Spectral) Analysis

Jean Baptiste Joseph Fourier (1768-1830)

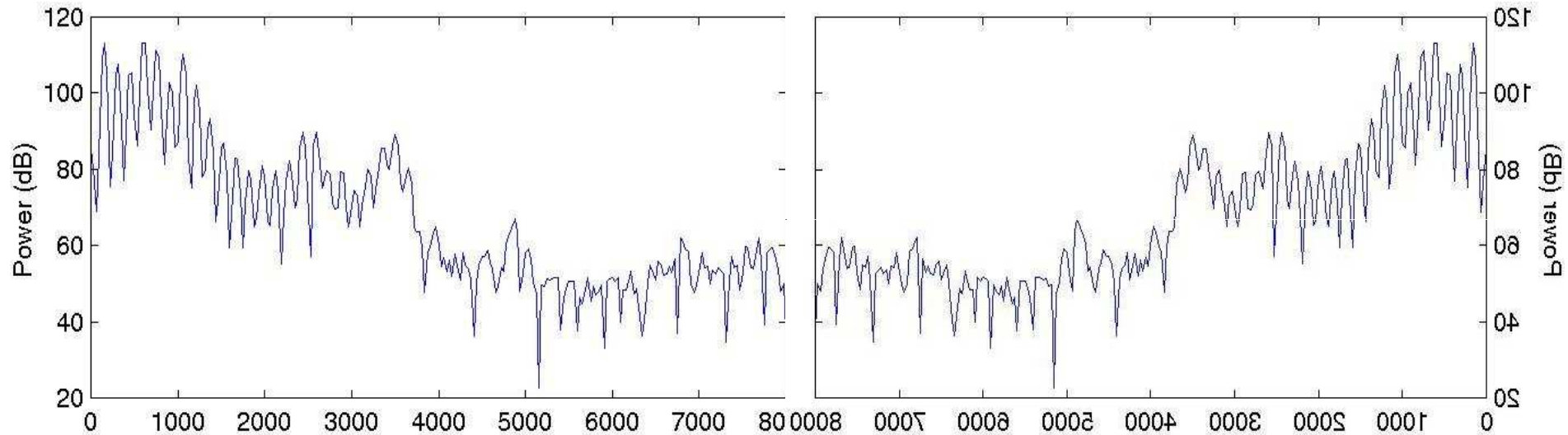


# Windowing



# Fast Fourier Transform

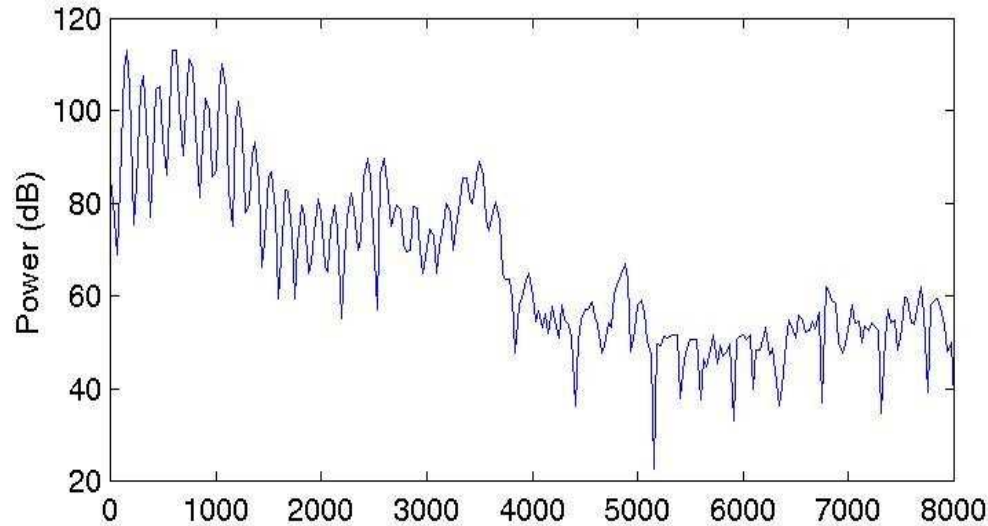
- Cooley and Tukey, mid 1960's
- e.g. for Power Spectrum



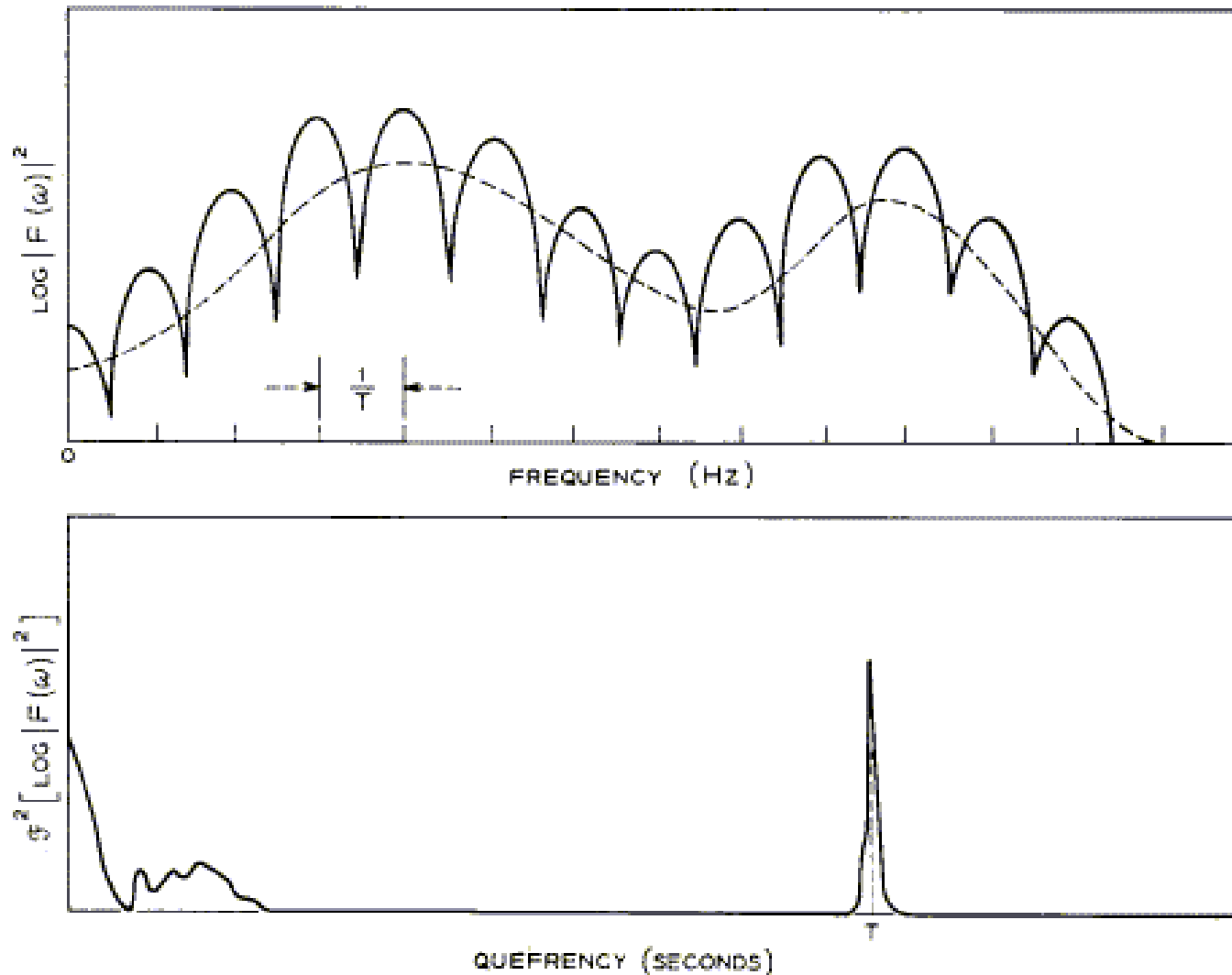


# Fast Fourier Transform

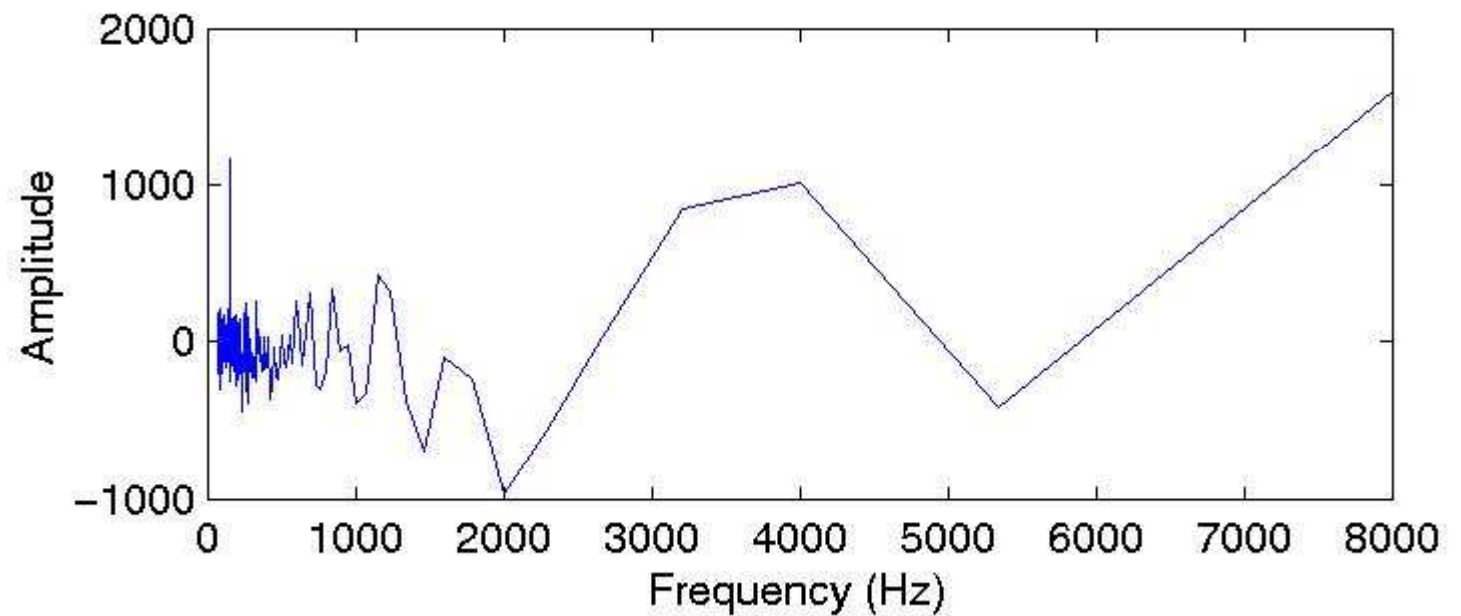
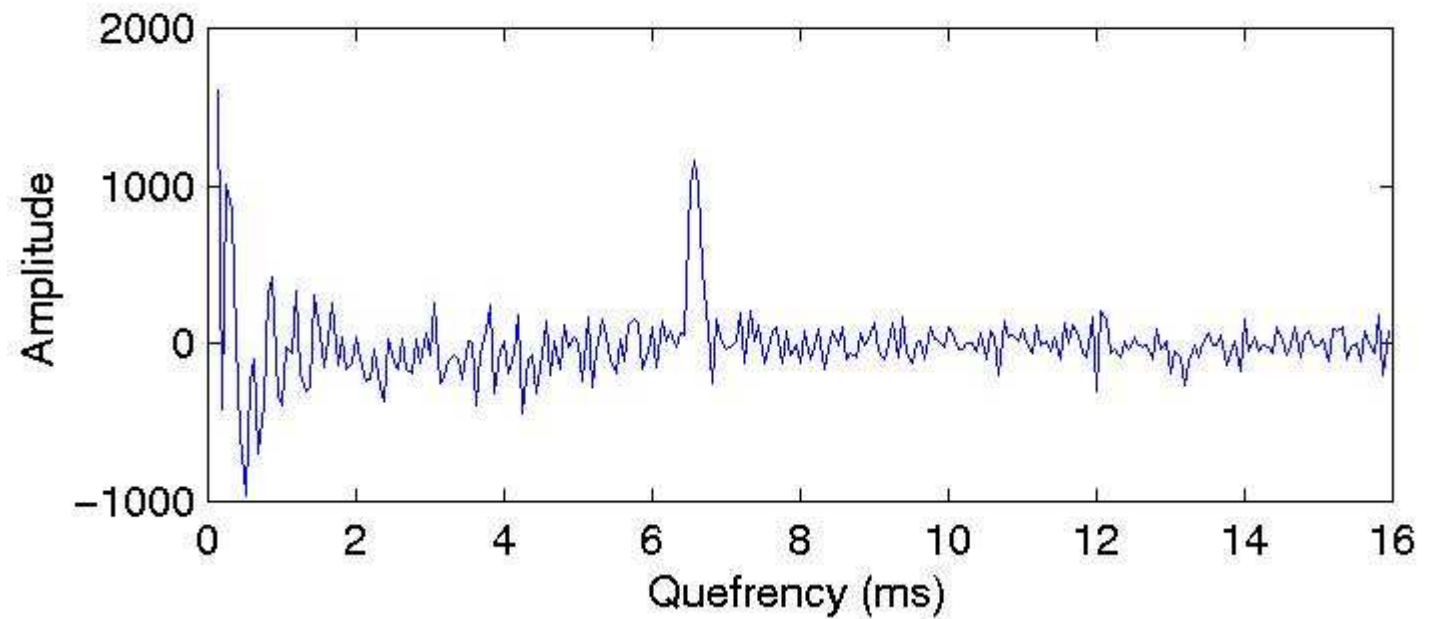
- Cooley and Tukey, mid 1960's
- e.g. for Power Spectrum



# Cepstrum (Noll 1967)

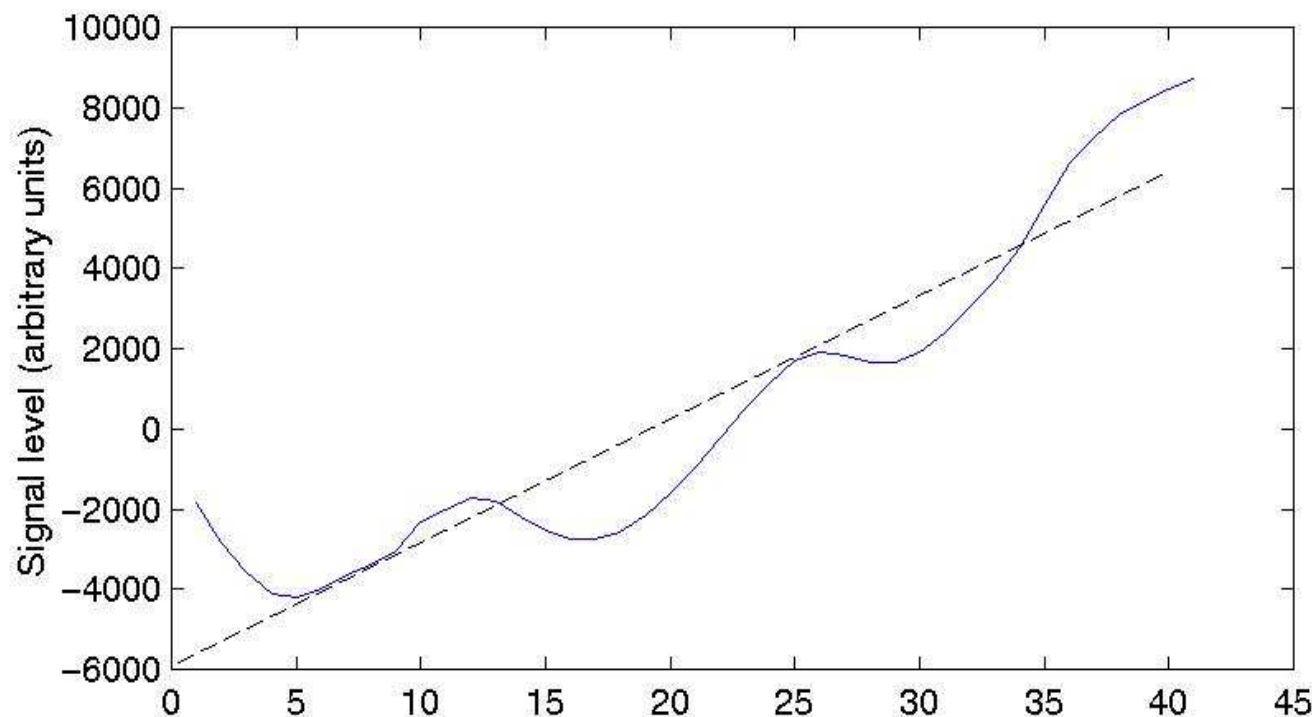


# Cepstrum

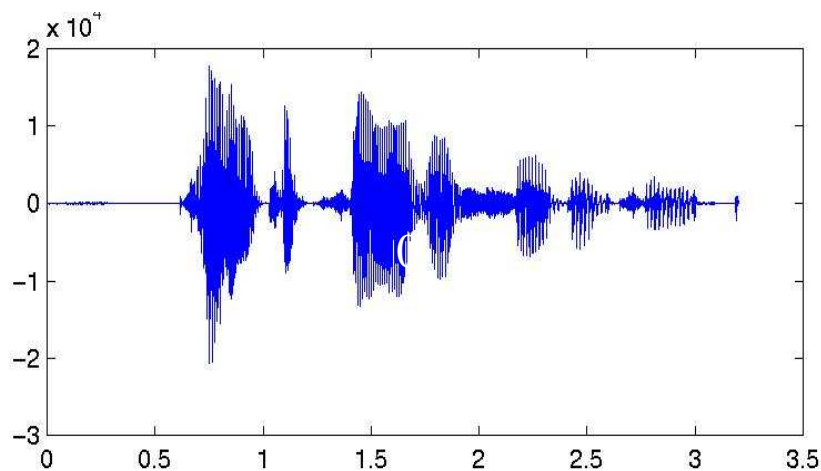


# Linear Prediction of speech

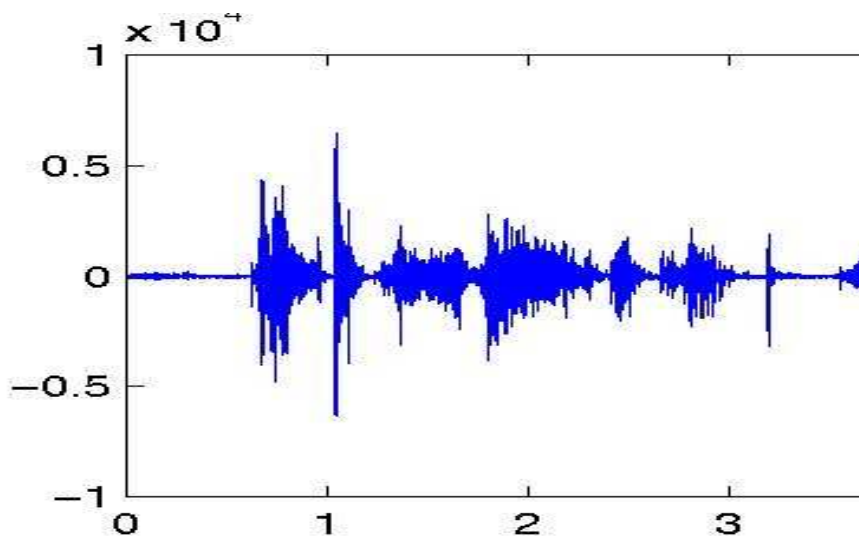
- We estimate the current sample as a linear combination of the previous  $p$  samples:
- $x[t] = -a_1 x[t-1] - a_2 x[t-2] - \dots - a_p x[t-p] + e[t]$



# Linear Prediction of speech

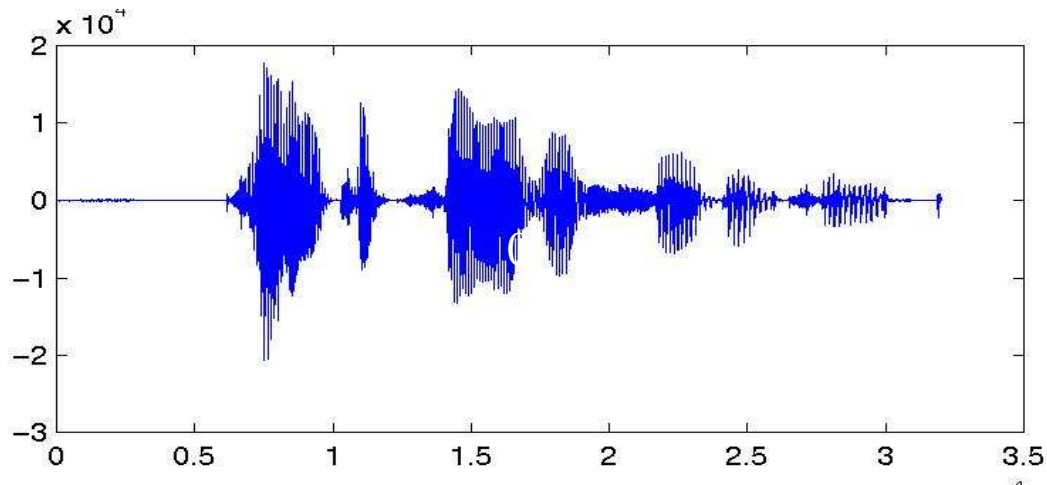


• Original

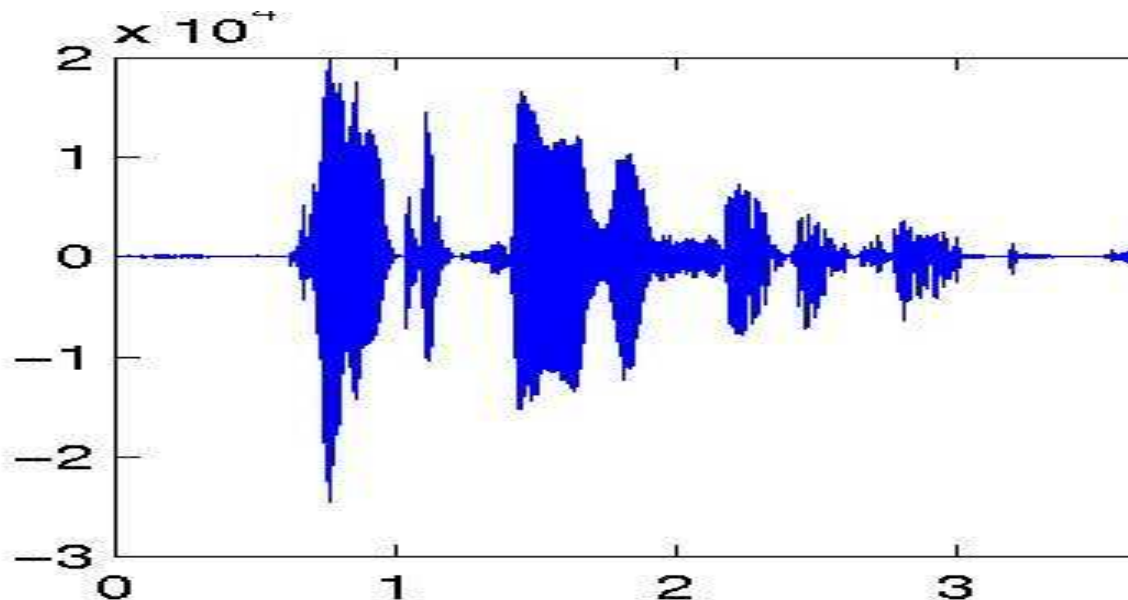


• Error

# Linear Prediction of speech

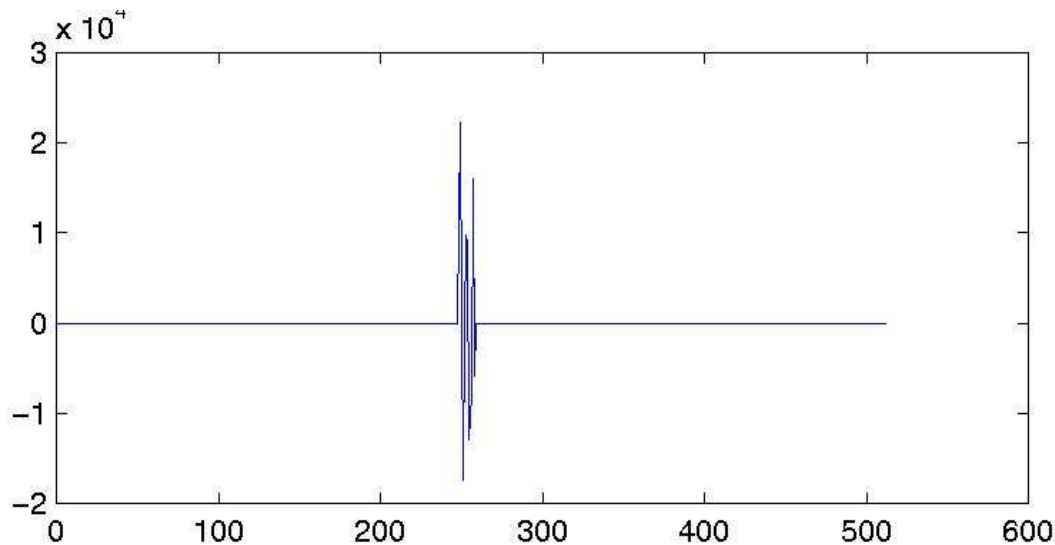
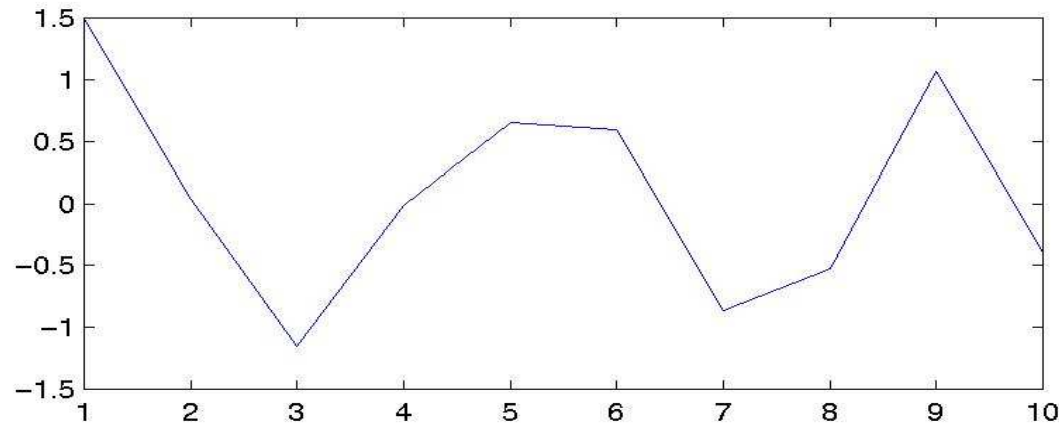


- Original



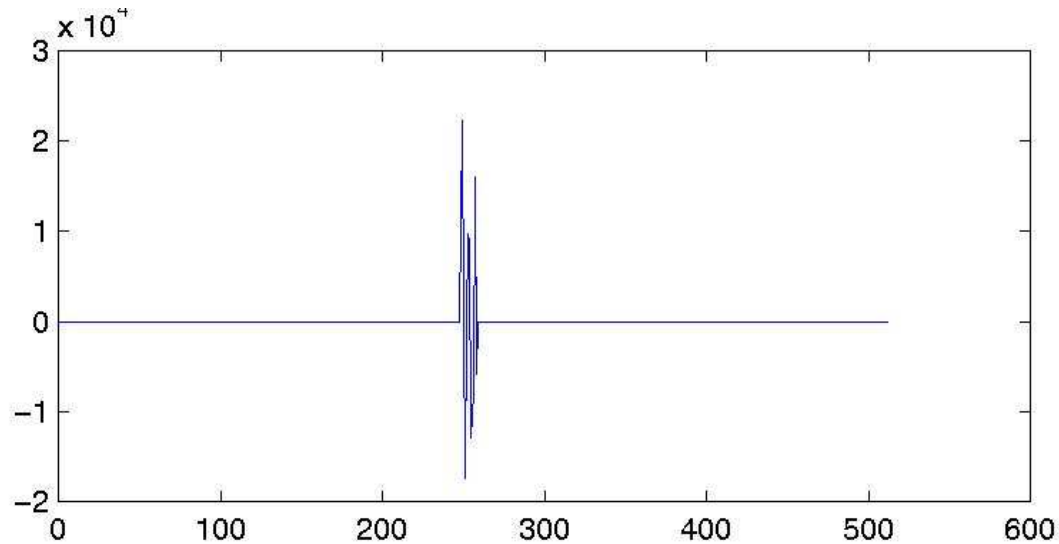
- Resynthesized  
excluding error

# A magic trick



- 10 LPC coefficients from an [a]
- Same 10 coeffs, scaled up and zero-padded to 512 samples long (251 each side)

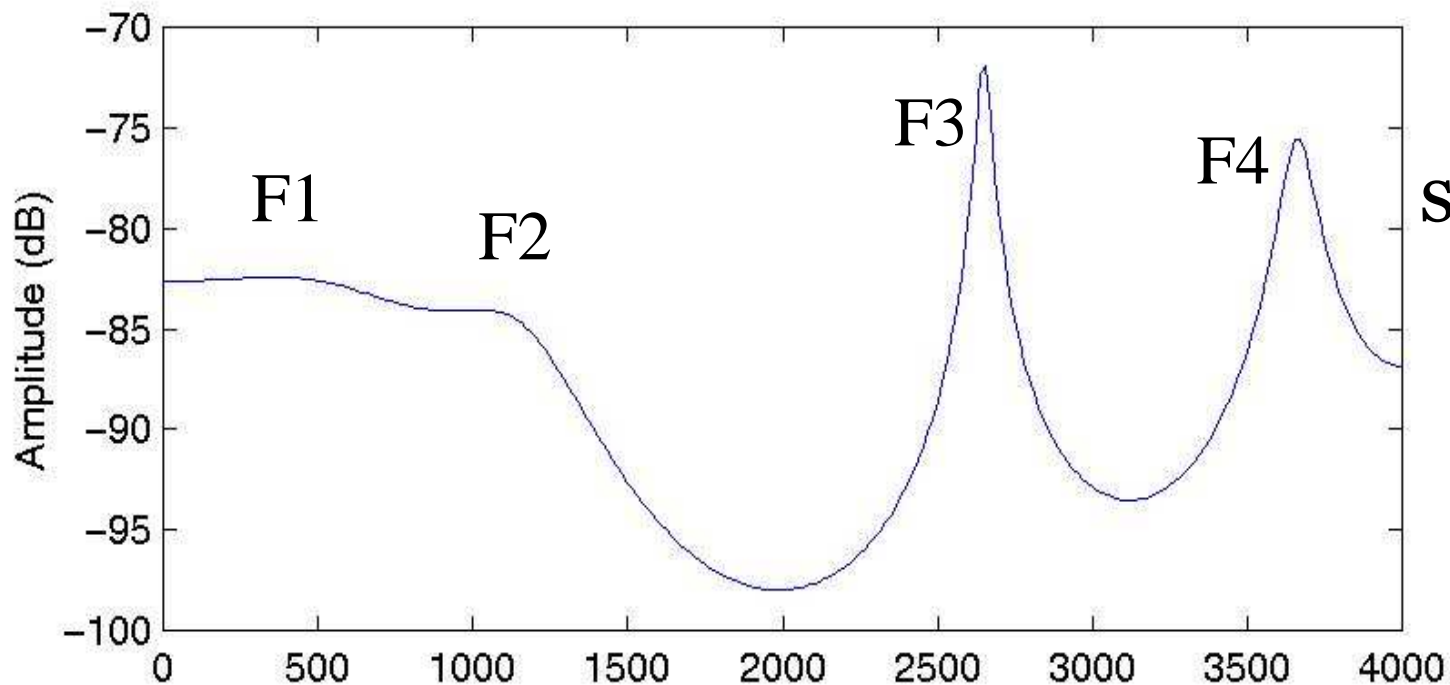
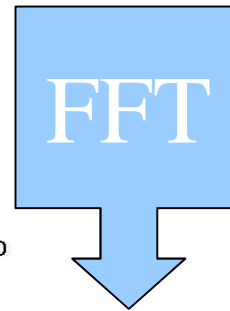
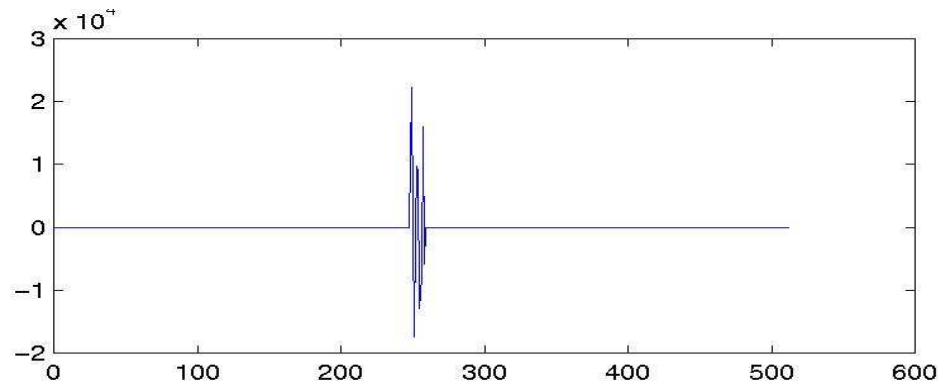
# A magic trick



- Same 10 coeffs, scaled up and zero-padded to 512 samples long



# A magic trick



LP  
spectrum

# Shameless plug

