



Spectral Analysis

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Joseph Fourier

lived from 1768 to 1830

Fourier studied the mathematical theory of heat conduction. He established the partial differential equation governing heat diffusion and solved it by using infinite series of trigonometric functions.


Find out more at:


<http://www-history.mcs.st-andrews.ac.uk/history/Mathematicians/Fourier.html>

[Spectral analysis 2](#)

Fourier transform

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$

 Spectrum at frequency f

 Basis function for frequency f

$$X(\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$

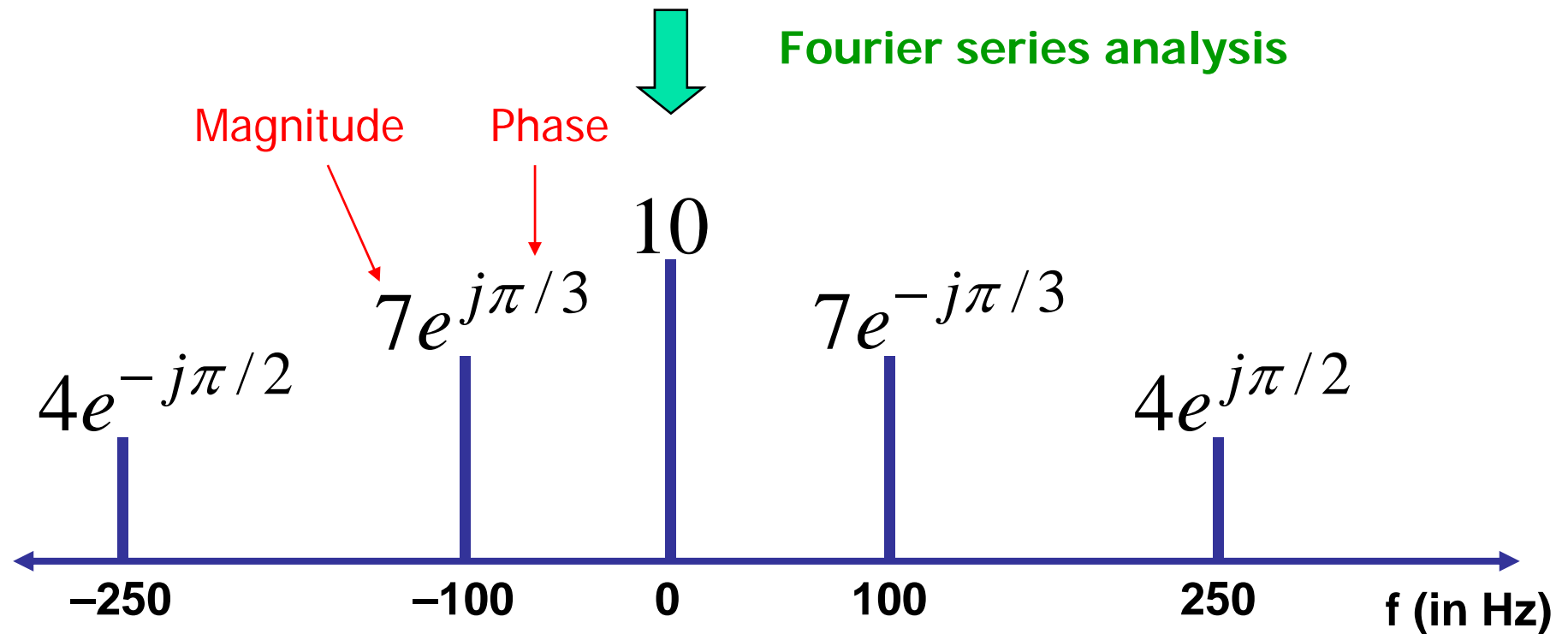
Inverse Fourier transform

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi ft} df$$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{j\omega t} d\omega$$

Example

$$x(t) = 10 + 14 \cos(2\pi(100)t - \pi/3) + 8 \cos(2\pi(250)t + \pi/2)$$



Magnitude and phase

$$X(f) = |X(f)| e^{j\phi(f)}$$

where magnitude and phase spectra

$$|X(f)| = \sqrt{\{\operatorname{Re}[X(f)]\}^2 + \{\operatorname{Im}[X(f)]\}^2}$$

$$\phi(f) = \tan^{-1} \left\{ \frac{\operatorname{Im}[X(f)]}{\operatorname{Re}[X(f)]} \right\}$$

% Generating two sinusoids and one DC component

```
fs=2000; % sampling at 2 kHz
t=0:1/fs:0.1;
x=10 + 14*cos(2*pi*100*t-pi/3) + 4*cos(2*pi*250*t-pi/2);
subplot(2,2,1)
plot(t,x)
ylabel('x(n)')
xlabel('Time, s')
title('Two sinusoids + DC')
axis([min(t) max(t) min(x)*1.1 max(x)*1.1])
```

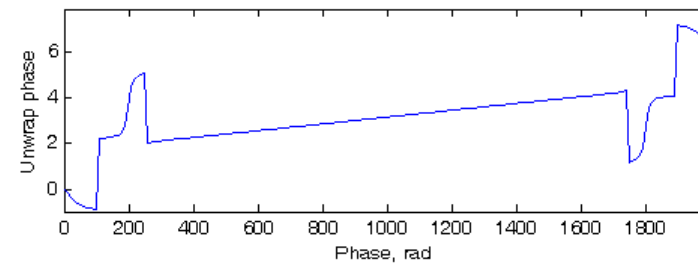
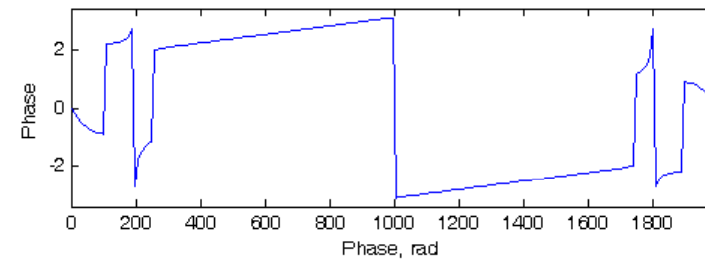
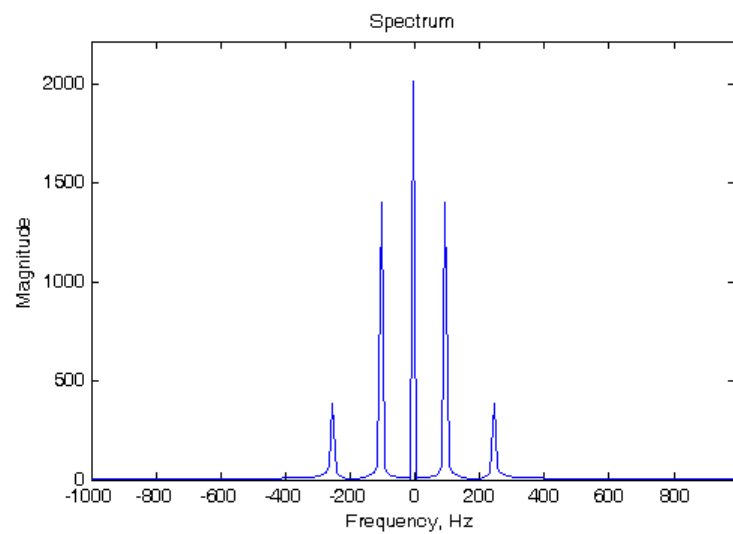
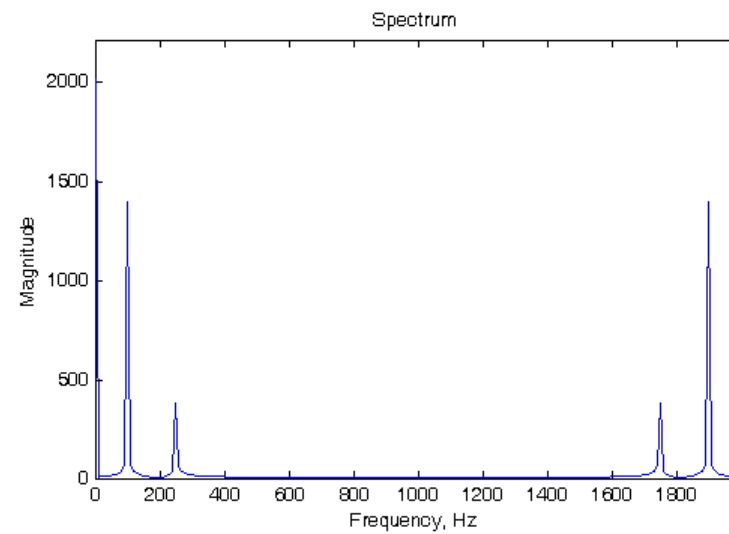
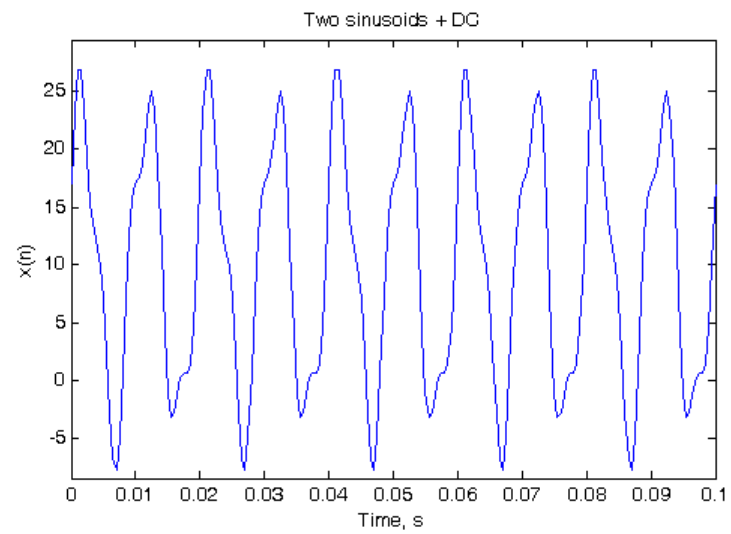
% Spectral analysis

```
Xf=fft(x);
resolution=fs/length(Xf);
f=(0:length(Xf)-1)*resolution;
Xf_mag = abs(Xf); % magnitude of spectrum
subplot(2,2,2)
plot(f,Xf_mag)
xlabel('Frequency, Hz')
ylabel('Magnitude')
title('Spectrum')
axis([min(f) max(f) 0 max(Xf_mag)*1.1])
```

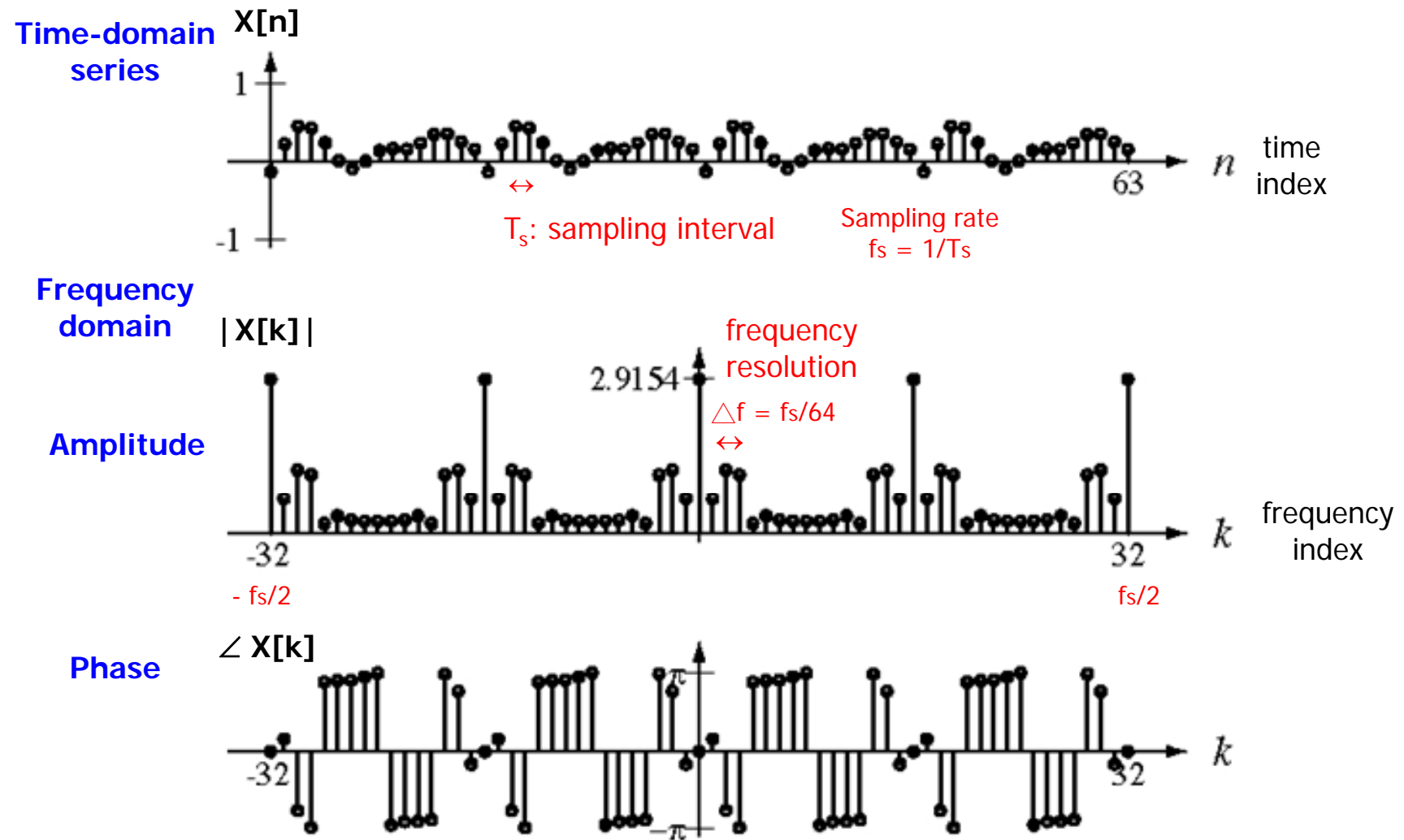
```
Xf_phase = angle(Xf);           % phase of spectrum
subplot(4,2,6)
plot(f,Xf_phase)
xlabel('Phase, rad'); ylabel('Phase');
axis([min(f) max(f) min(Xf_phase)*1.1 max(Xf_phase)*1.1])
```

```
Xf_phase = unwrap(Xf_phase);    % Unwrap phase angle
subplot(4,2,8)
plot(f,Xf_phase)
xlabel('Phase, rad'); ylabel('Unwrap phase');
axis([min(f) max(f) min(Xf_phase)*1.1 max(Xf_phase)*1.1])
```

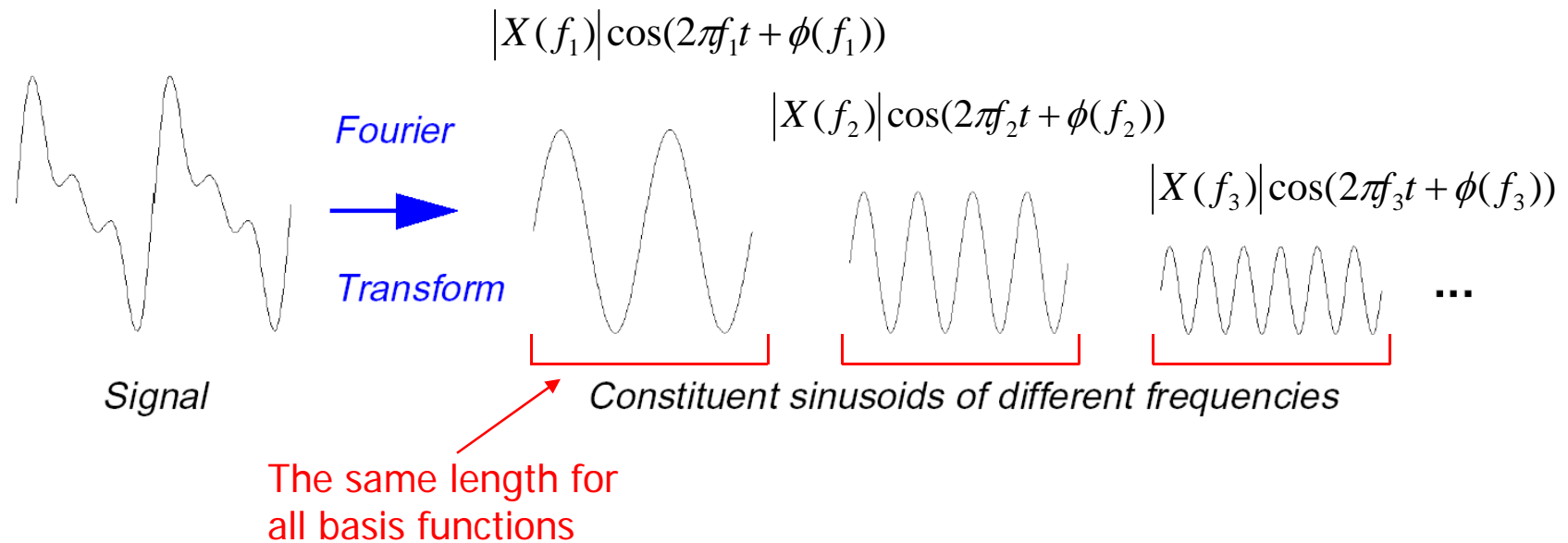
```
% Shift zero-frequency component to center of spectrum
Xf_center=fftshift(Xf);
Xf_mag = abs(Xf_center);        % magnitude of spectrum
Xf_phase = phase(Xf_center);    % phase of spectrum
subplot(2,2,3)
f=f-fs/2;
plot(f,Xf_mag)
xlabel('Frequency, Hz'); ylabel('Magnitude');
title('Spectrum')
axis([min(f) max(f) 0 max(Xf_mag)*1.1])
```



Discrete Fourier transform



Decomposition of signal



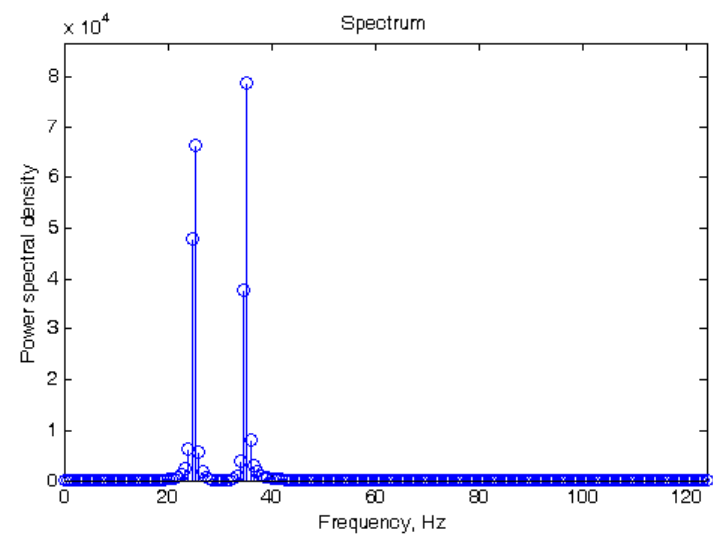
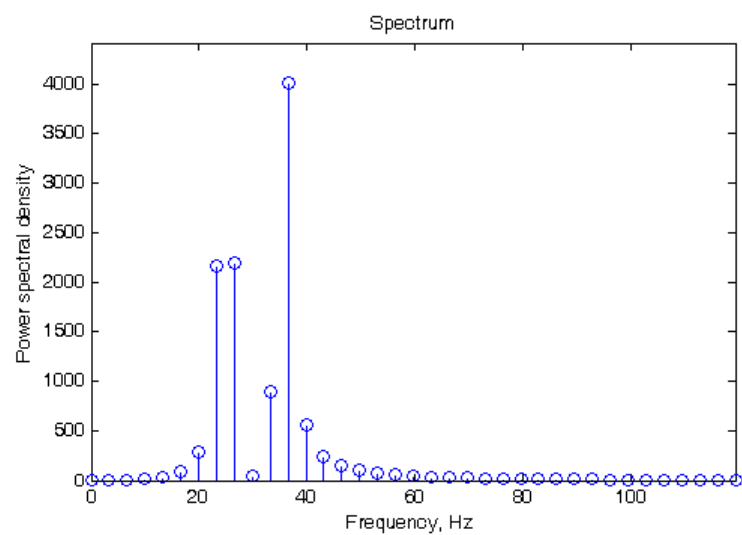
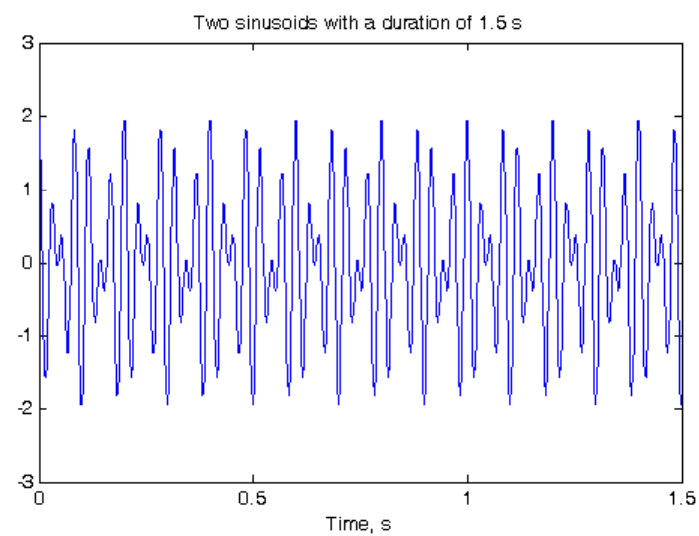
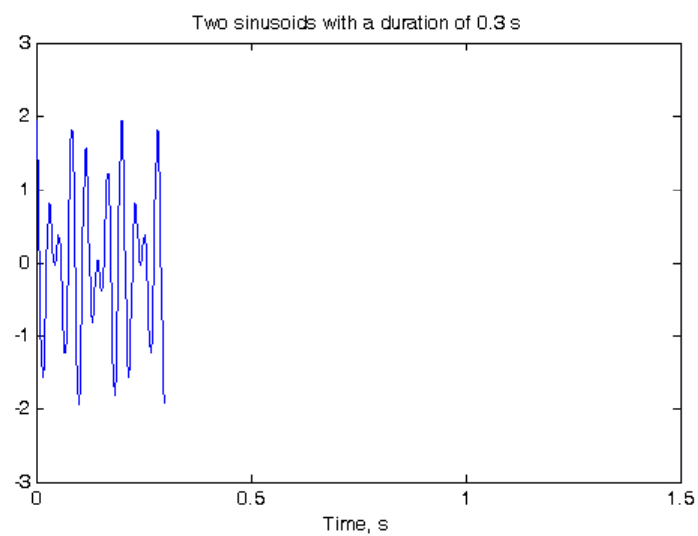
Modified from "Matlab Wavelet Toolbox Manual"

Effect of data points on spectral analysis

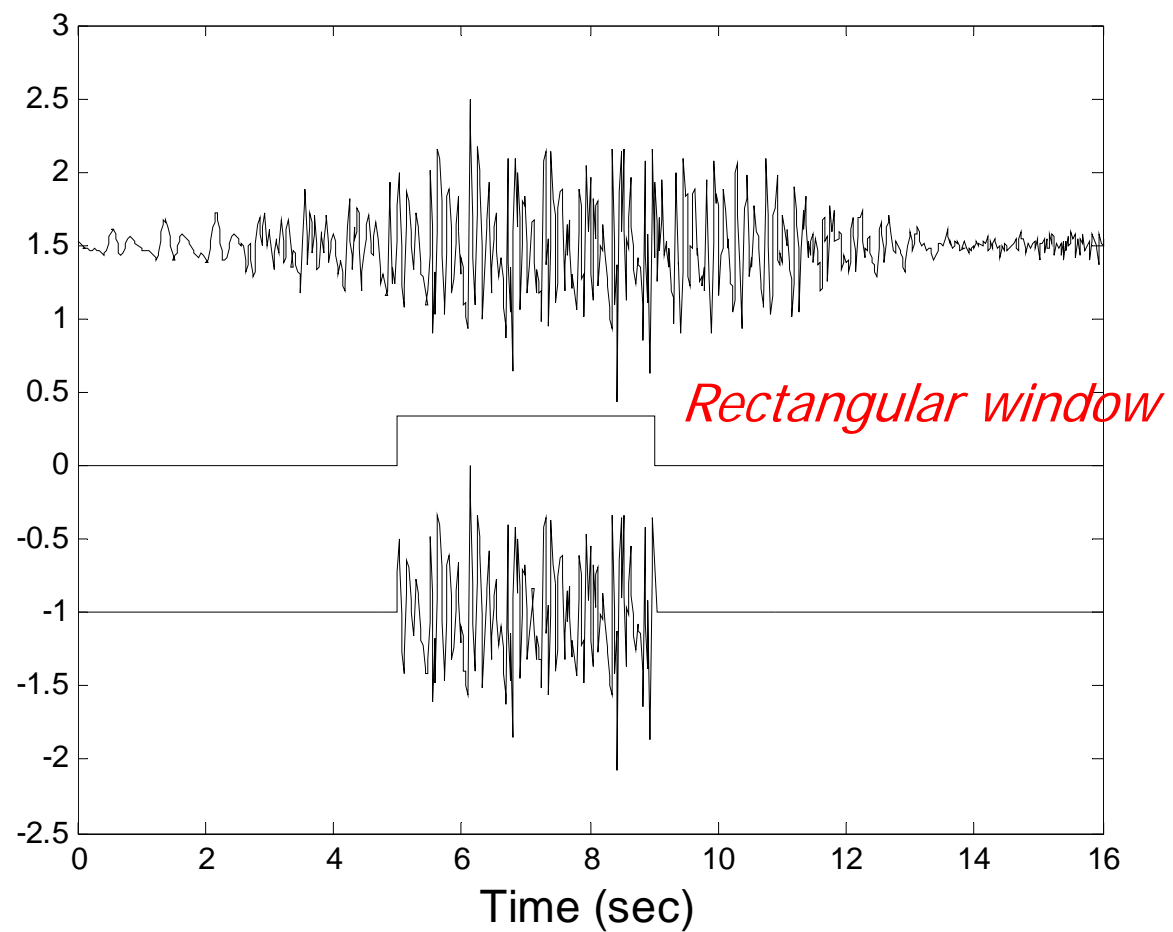
```
fs=500;  
t=0:1/fs:0.3;  
x = cos(2*pi*25*t) + cos(2*pi*35*t+pi/10);  
  
Xf=fft(x);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
index=1:length(Xf)/4;  
stem(f(index),Xf_power(index))
```

% Improve resolution by increase data length

```
t=0:1/fs:0.3*5;  
x = cos(2*pi*25*t) + cos(2*pi*35*t+pi/10);  
  
Xf=fft(x);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
index=1:length(Xf)/4;  
stem(f(index),Xf_power(index))
```



Data length: truncation



Window functions

Rectangular:

$$w(n) = 1$$

Blackman:

$$w(n) = 0.42 - 0.5 \cos\left(\frac{2\pi n}{N-1}\right) + 0.08 \cos\left(\frac{4\pi n}{N-1}\right)$$

Hamming:

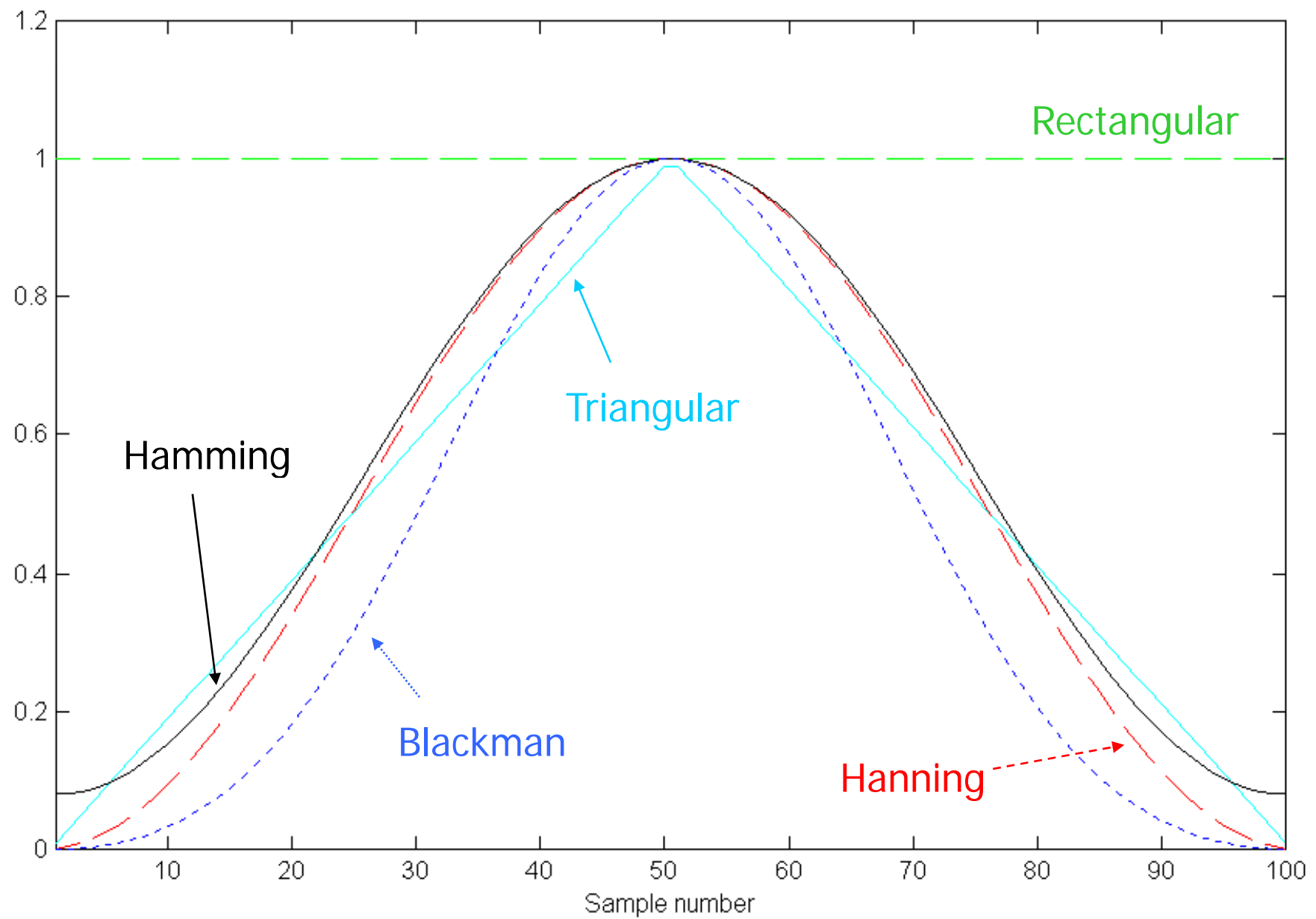
$$w(n) = 0.54 - 0.46 \cos\left(\frac{2\pi n}{N-1}\right)$$

Bartlett (Triangular):

$$w(n) = \begin{cases} \frac{2n}{N-1}, & 0 \leq n \leq \frac{N-1}{2} \\ 2 - \frac{2n}{N-1}, & \frac{N-1}{2} \leq n \leq N-1 \end{cases}$$

Hanning:

$$w(n) = 0.5 - 0.5 \cos\left(\frac{2\pi n}{N-1}\right)$$



```
fs=500;  
t=0:1/fs:0.3;  
x = cos(2*pi*25*t) + cos(2*pi*35*t+pi/10);
```

```
Xf=fft(x);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
index=1:length(Xf)/4;  
stem(f(index),Xf_power(index))
```

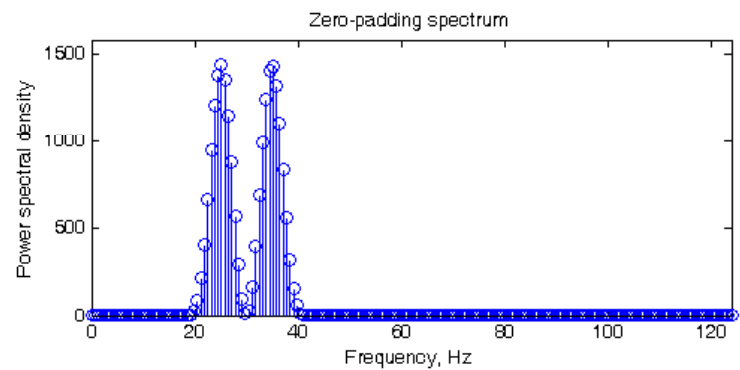
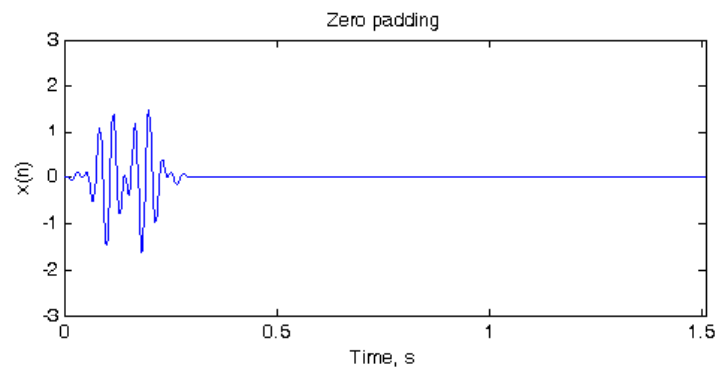
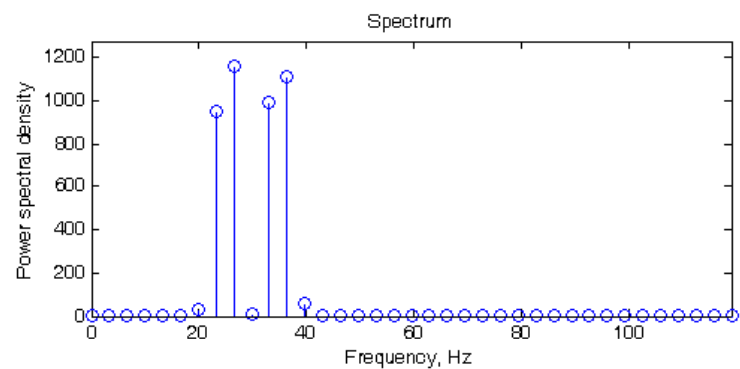
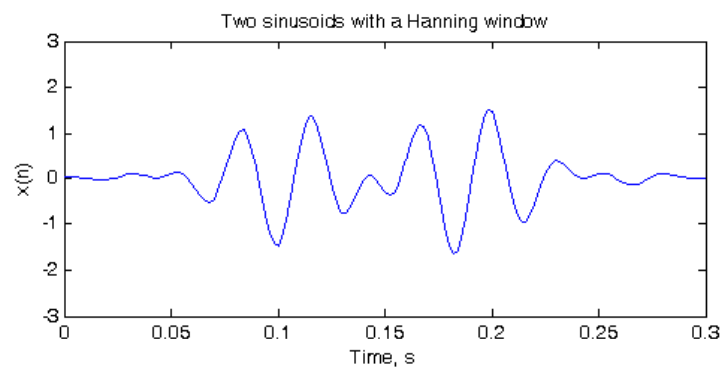
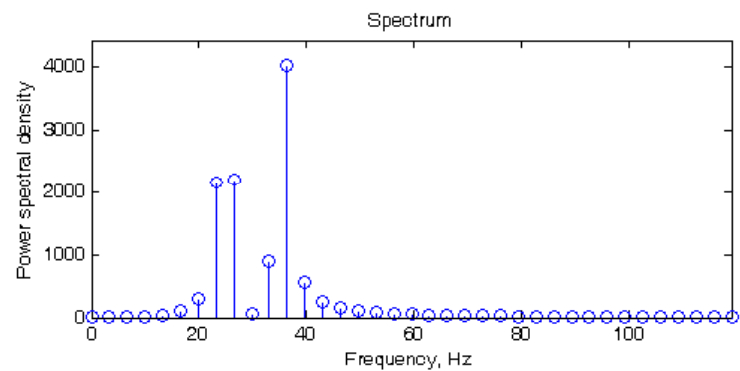
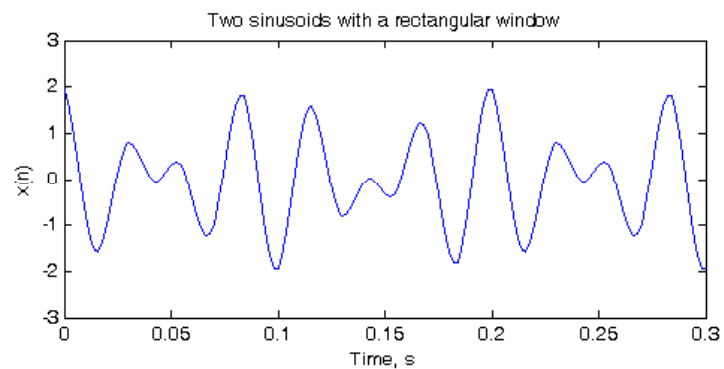
% Using Hanning window

```
x=x.*hanning(length(x))';  
Xf=fft(x);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
index=1:length(Xf)/4;  
stem(f(index),Xf_power(index))
```

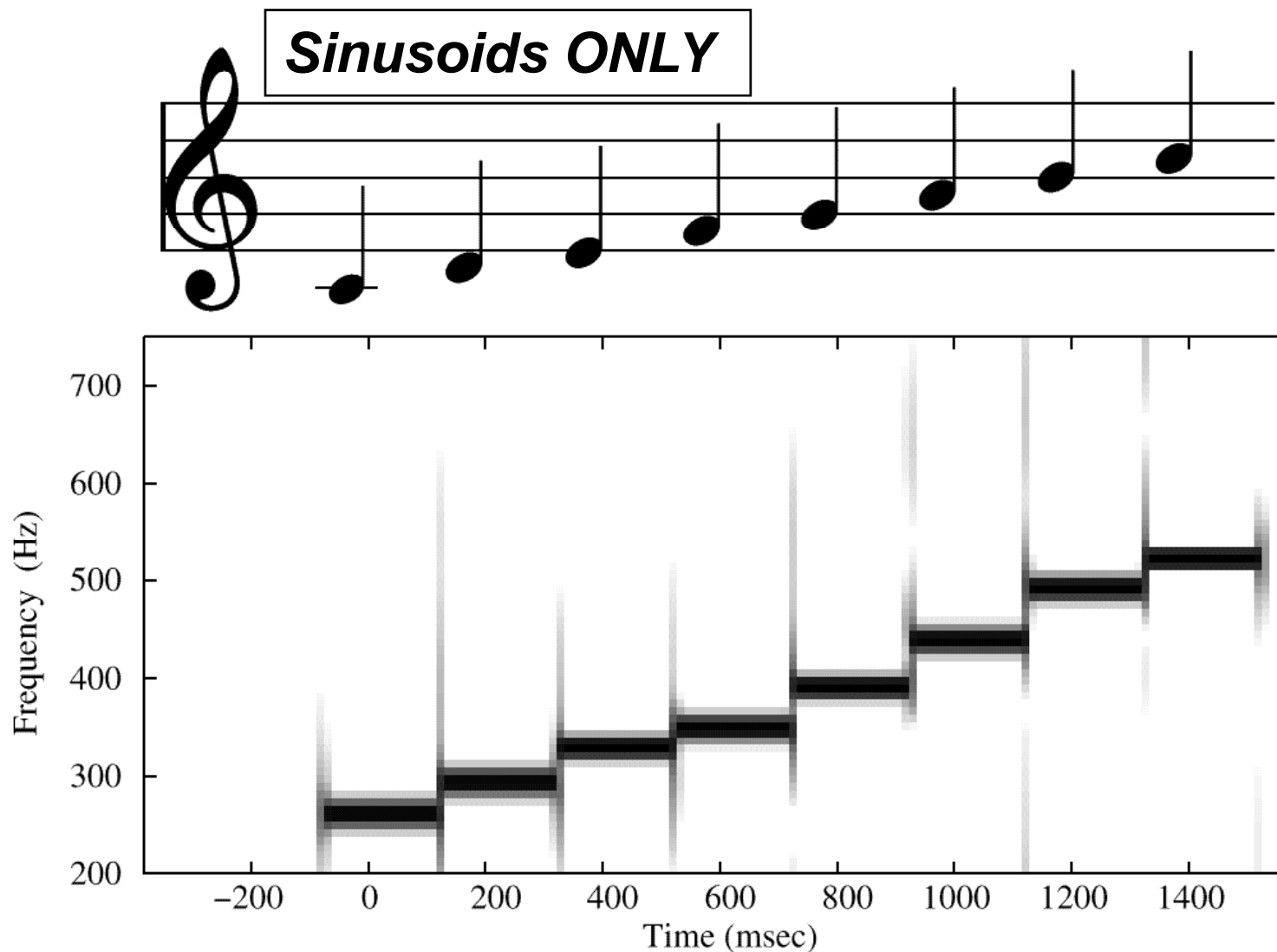
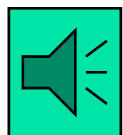

`% Zero padding`

```
x_ZP=zeros(1,length(x)*5);  
x_ZP(1:length(x))=x;  
t_ZP=(0:length(x_ZP)-1)*1/fs;
```

```
Xf=fft(x,length(x)*5);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
index=1:length(Xf)/4;  
stem(f(index),Xf_power(index))
```

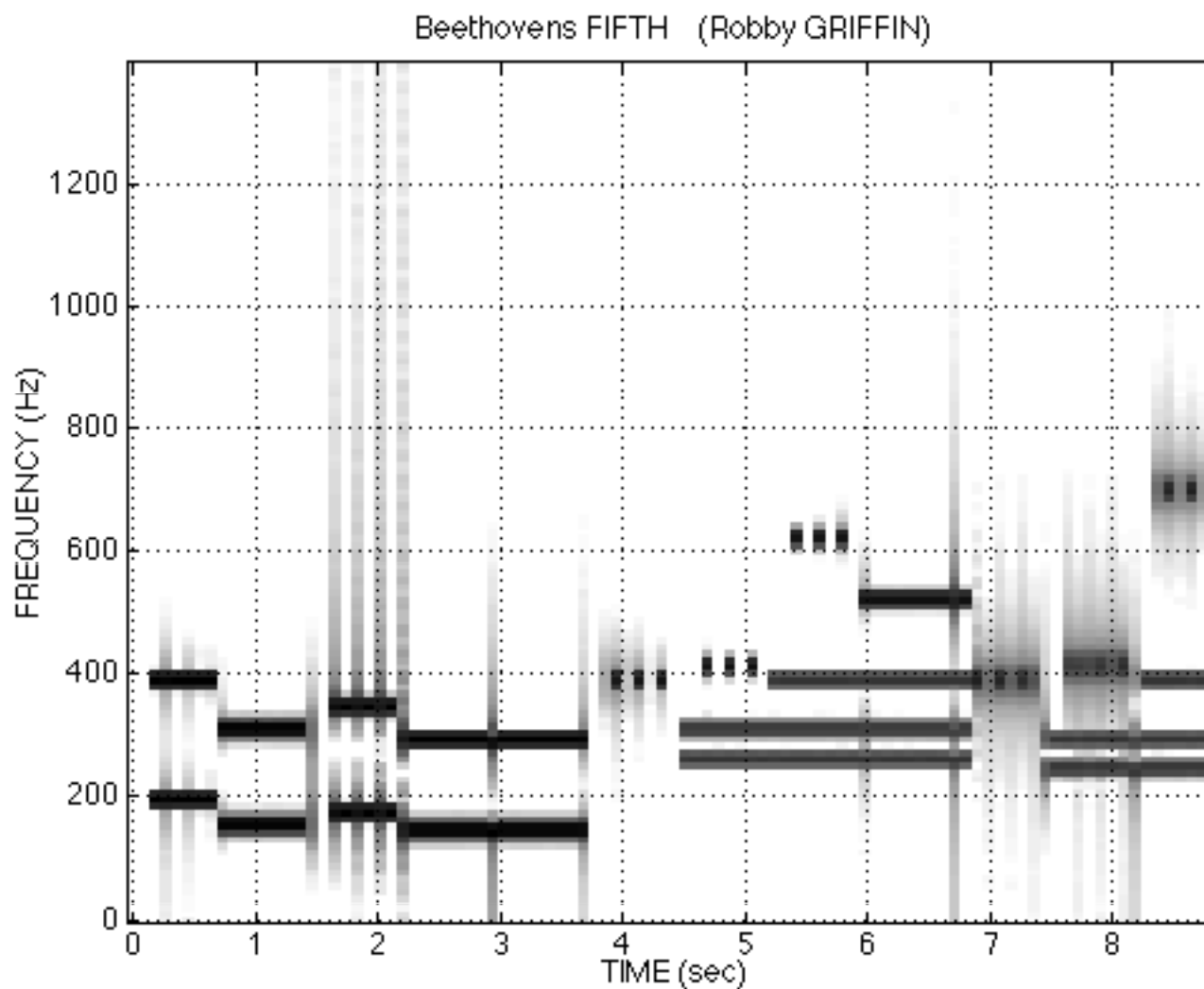
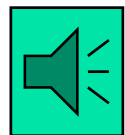


Spectrogram of C-Scale



From J.H. McClellan, R.W. Schafer, Signal Processing First, Prentice-Hall, 2003.

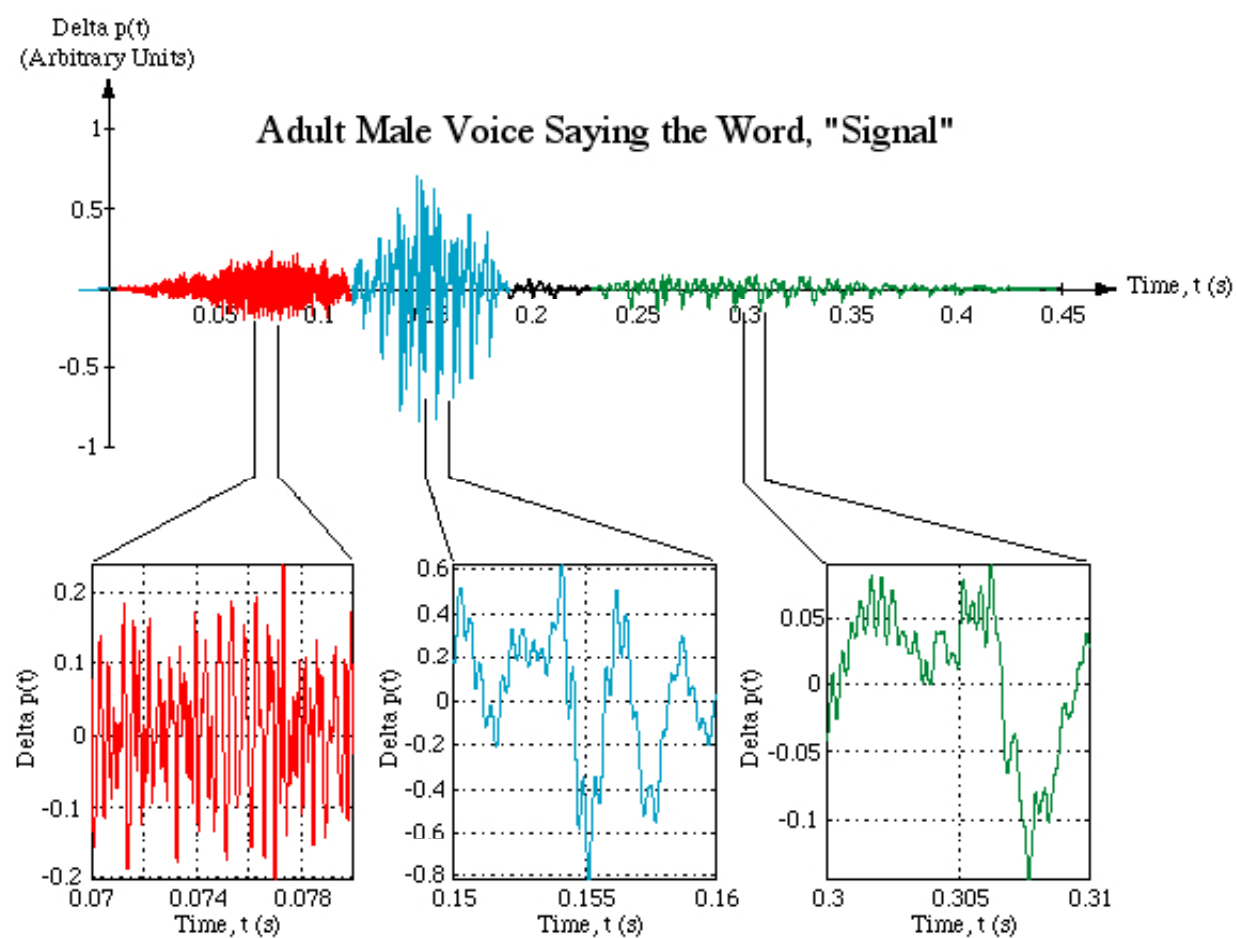
Spectrogram of LAB SONG



From J.H. McClellan, R.W. Schafer, Signal Processing First, Prentice-Hall, 2003.

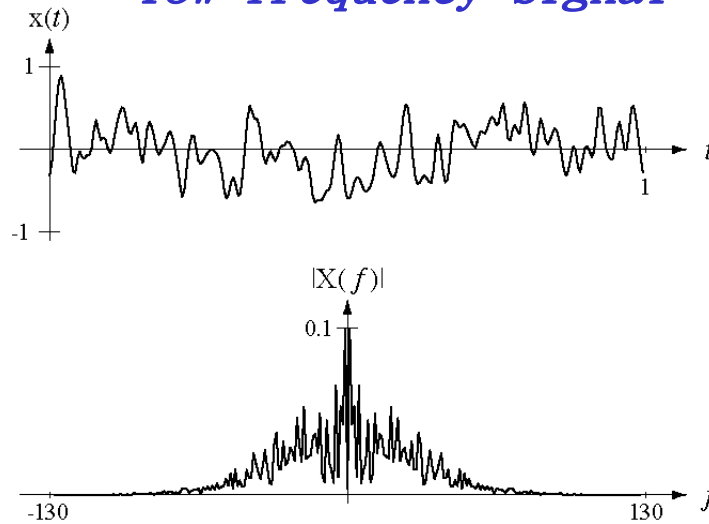
Recorded sound

“s” “i” “gn” “al”

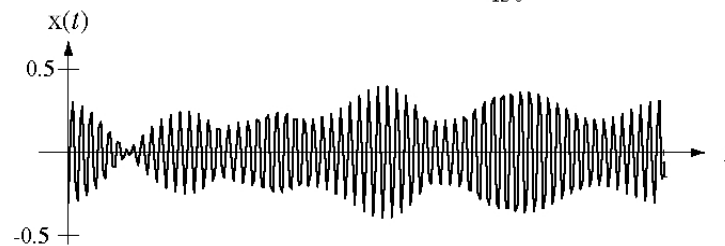
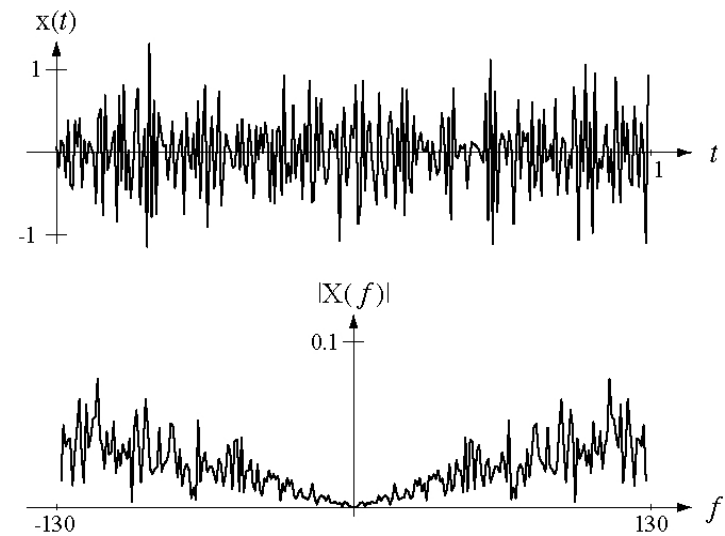


Fourier transform of signals

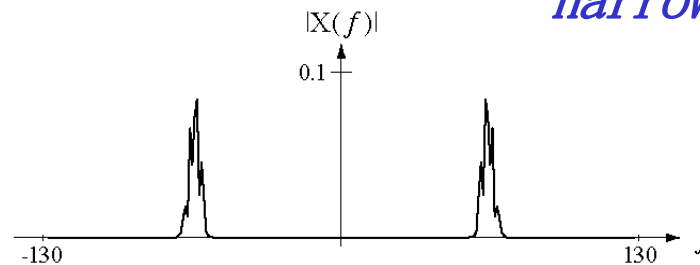
low-frequency signal



high-frequency signal



narrow-band signal

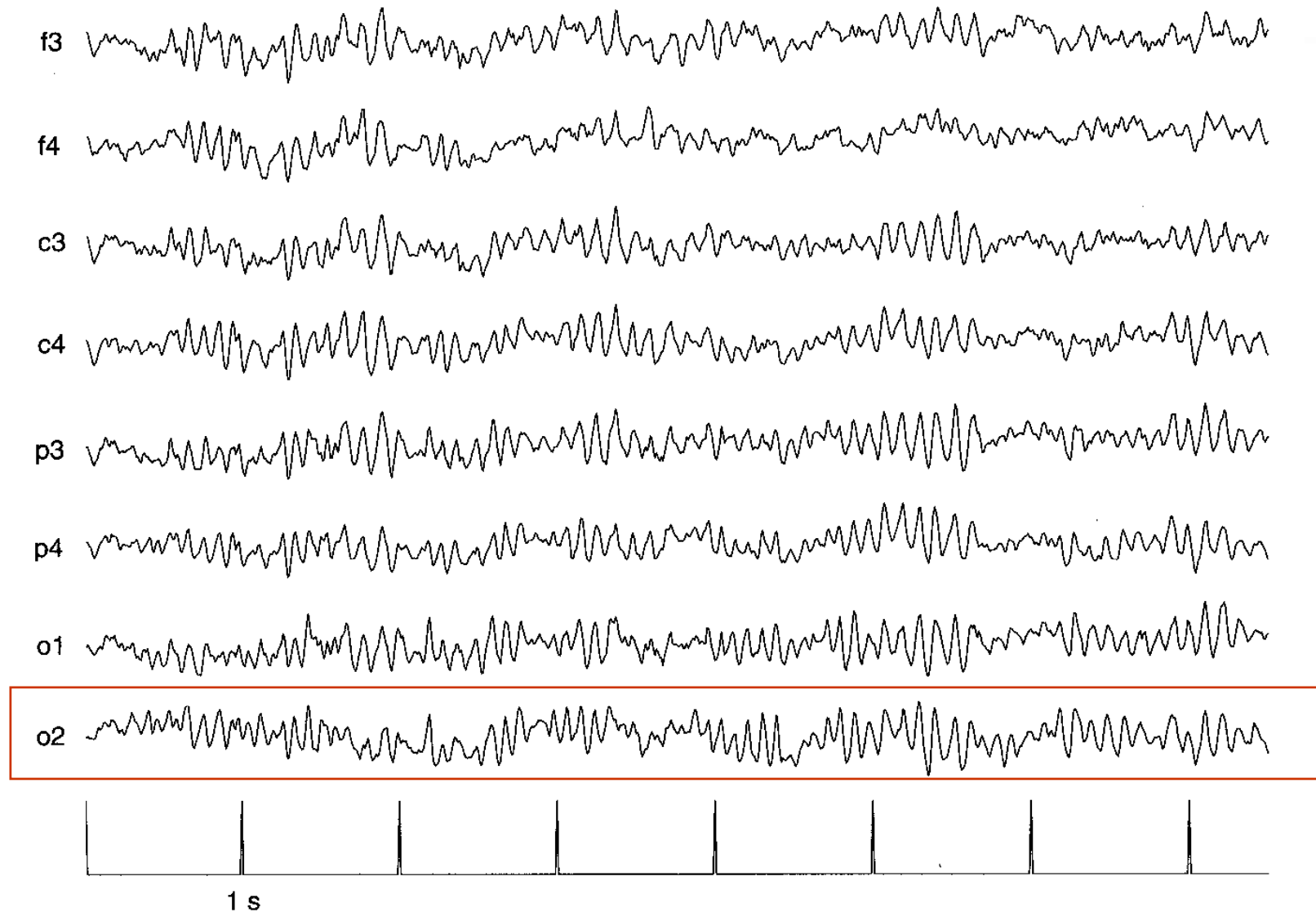


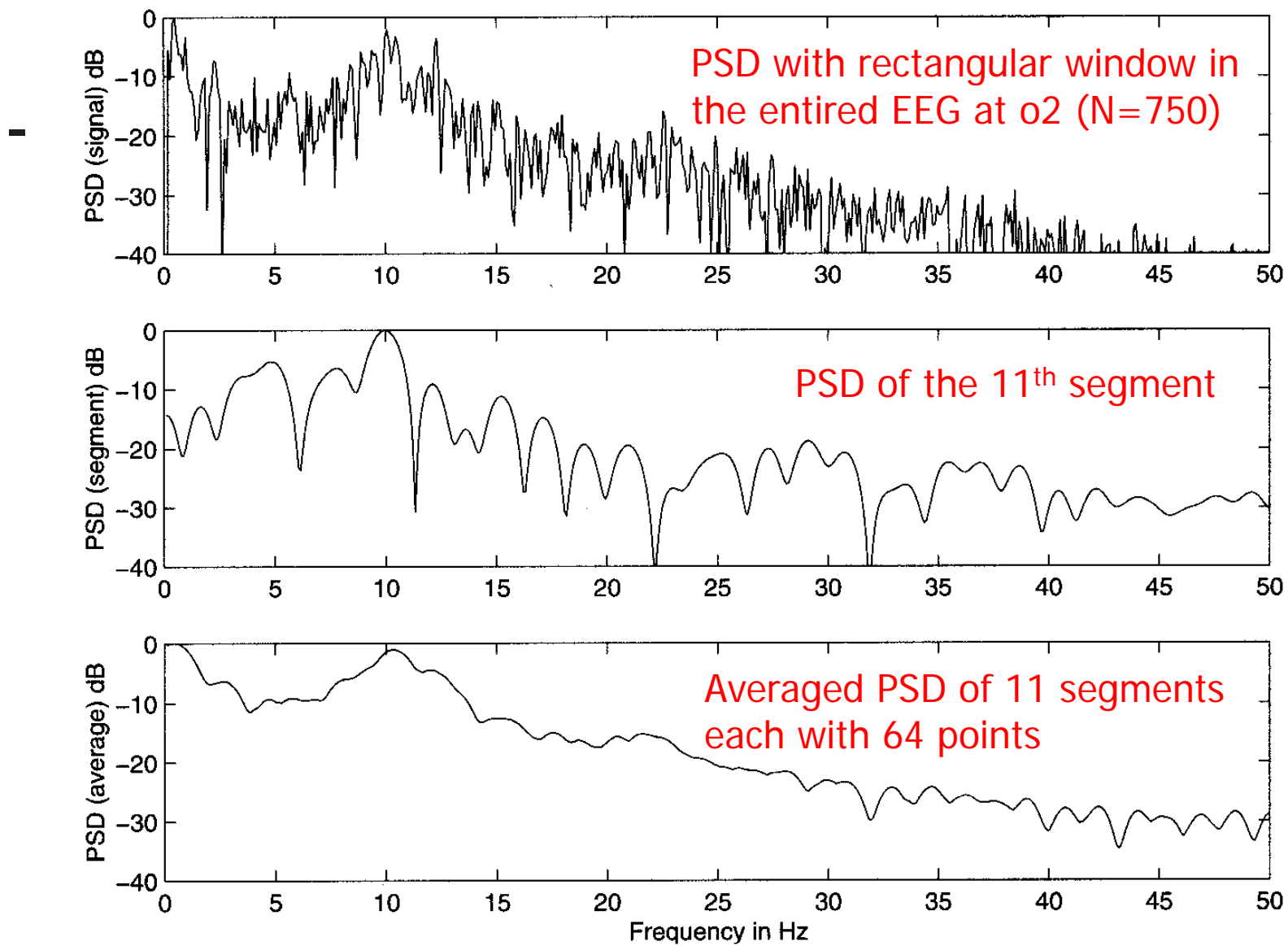
From .J.Roberts, Signals and Systems, M
cGraw-Hill, 2003.

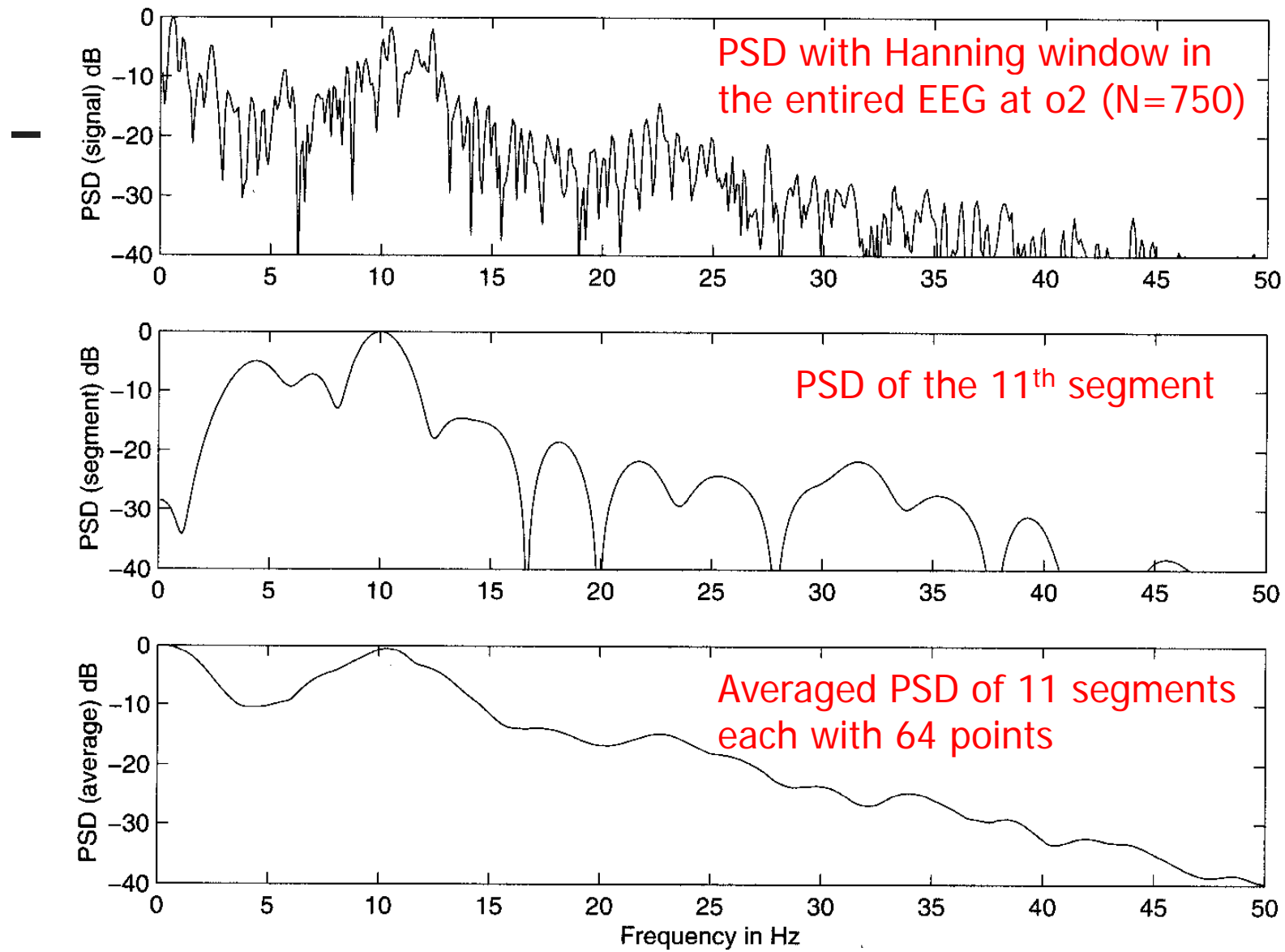
Averaged periodogram

- A signal is divided into non-overlapping segments
 - Reducing variance of spectral analysis at the sacrifice with frequency resolution
- Welch periodogram
 - Using overlapping segments.
 - Widely use in spectral estimation

Characterization of EEG Alpha Rhythms



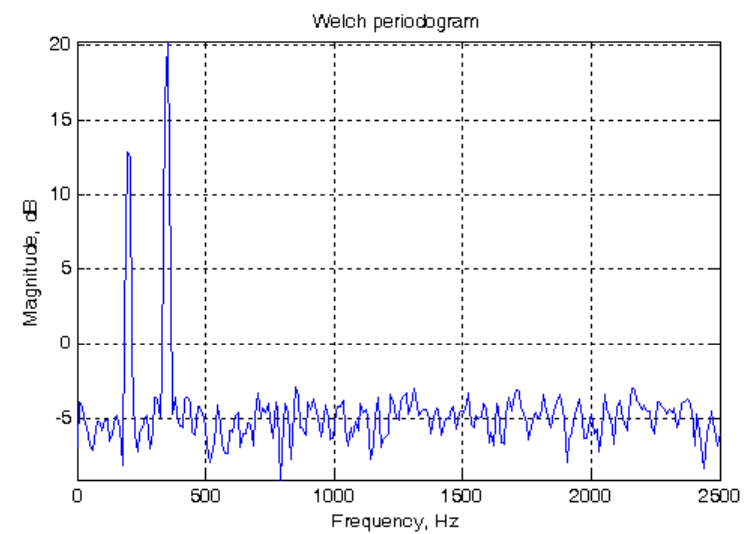
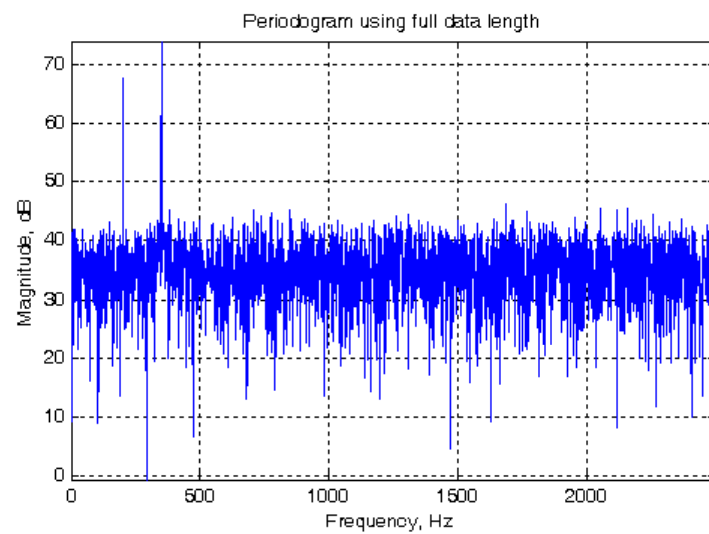
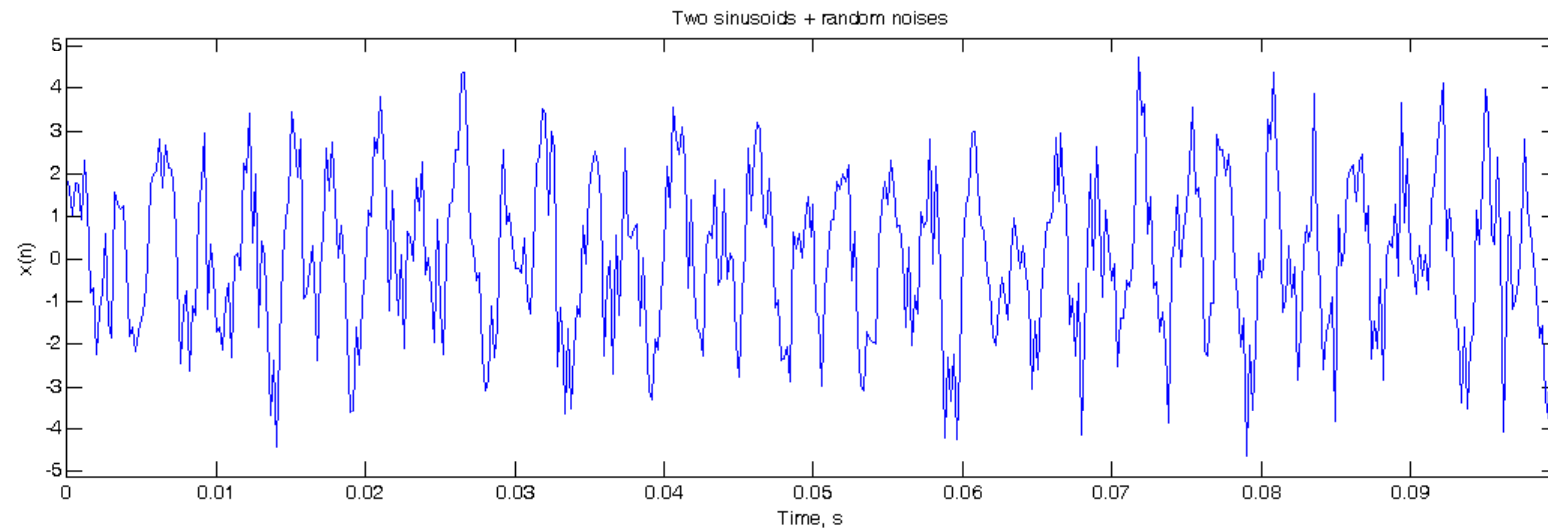




```
fs=5000;  
t=0:1/fs:1;  
x = sin(2*pi*200*t) + 2*sin(2*pi*350*t) + randn(size(t));
```

```
Xf=fft(x);  
resolution=fs/length(Xf);  
f=(0:length(Xf)-1)*resolution;  
Xf_power = Xf.*conj(Xf); % power spectral density  
Xf_dB=10*log10(Xf_power);  
index=1:length(Xf_power)/2;  
plot(f(index),Xf_dB(index))
```

```
NFFT=512; % length of window  
[P1,f] = pwelch(x,hanning(NFFT),NFFT/2,NFFT);  
% PWELCH(X,WINDOW,NOVERLAP,NFFT)  
f=f/pi*fs/2;  
P1_dB=10*log10(P1);  
plot(f,P1_dB)
```



Cross spectral density and coherence

Cross spectrum

$$S_{xy}(f) = X(f) \cdot Y^*(f)$$

Transfer function

$$H(f) = Y(f) / X(f)$$

Coherence function

$$\gamma_{xy}^2(f) = \frac{\left| \frac{1}{K} \sum_{k=1}^K X_k(f) Y_k^*(f) \right|^2}{\frac{1}{K} \sum_{k=0}^K X_k(f) \cdot X_k^*(f) \times \frac{1}{K} \sum_{k=0}^K Y_k(f) \cdot Y_k^*(f)}$$

$$\text{where } 0 \leq \gamma_{xy}^2(f) \leq 1$$

% Generate two sinudoids with the same frequency

```
fs=5000;  
t=0:1/fs:1;  
x = sin(2*pi*500*t) + randn(size(t));  
y = sin(2*pi*500*t+pi/10) + randn(size(t));
```

% Welch periodogram

```
NFFT=512;  
[P1,f] = pwelch(x,hanning(NFFT),NFFT/2,NFFT);  
f=f/pi*fs/2;  
plot(f,P1)  
[P2,f] = pwelch(y,hanning(NFFT),NFFT/2,NFFT);  
f=f/pi*fs/2;  
plot(f,P2)
```

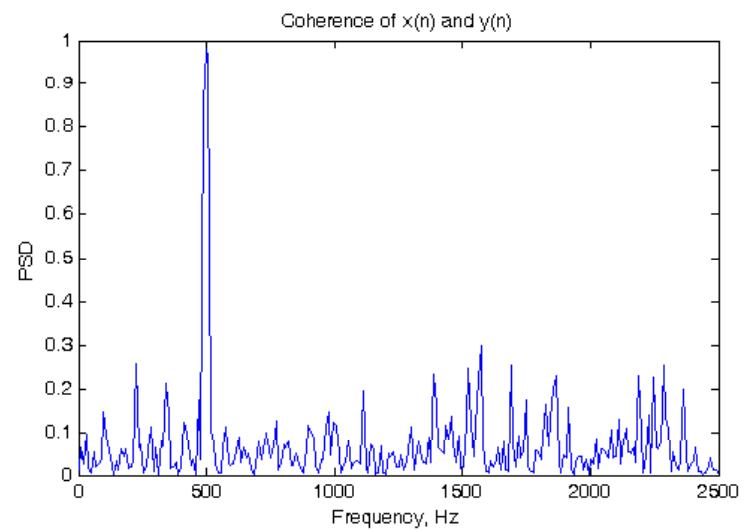
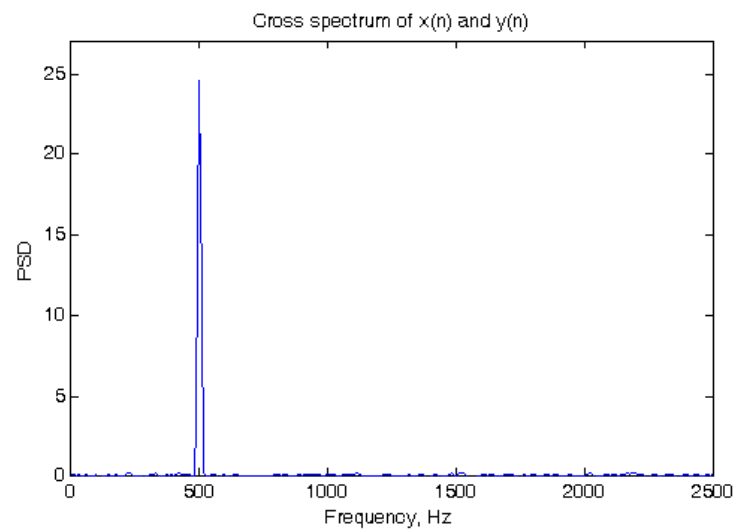
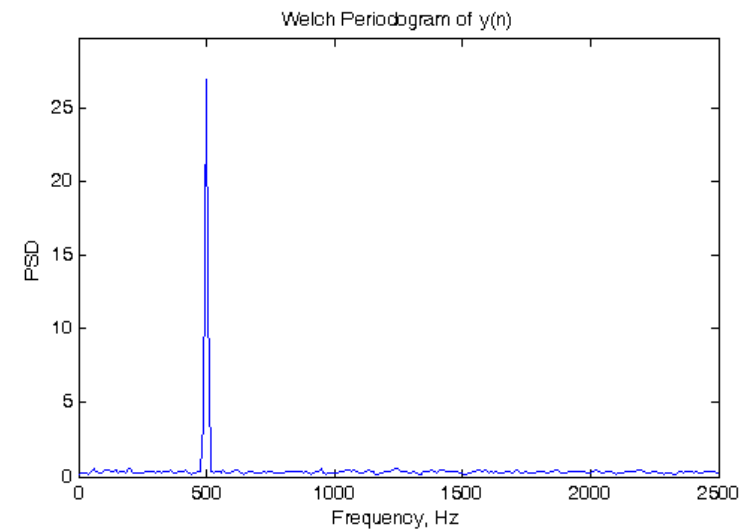
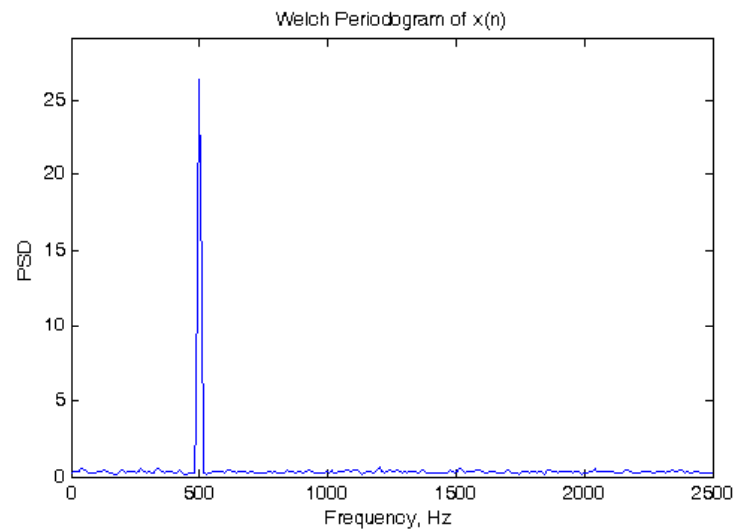
% Cross spectral analysis

```
[Pxy,f] = cpsd(x,y,hanning(NFFT),NFFT/2,NFFT);  
f=f/pi*fs/2;  
plot(f,Pxy)
```

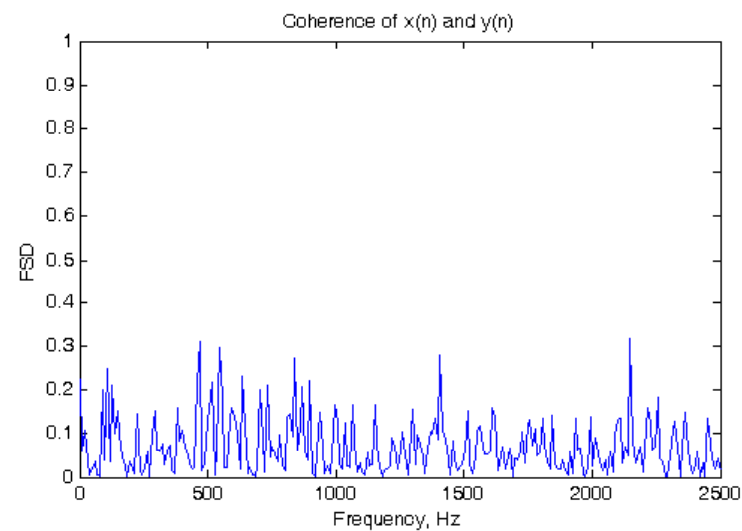
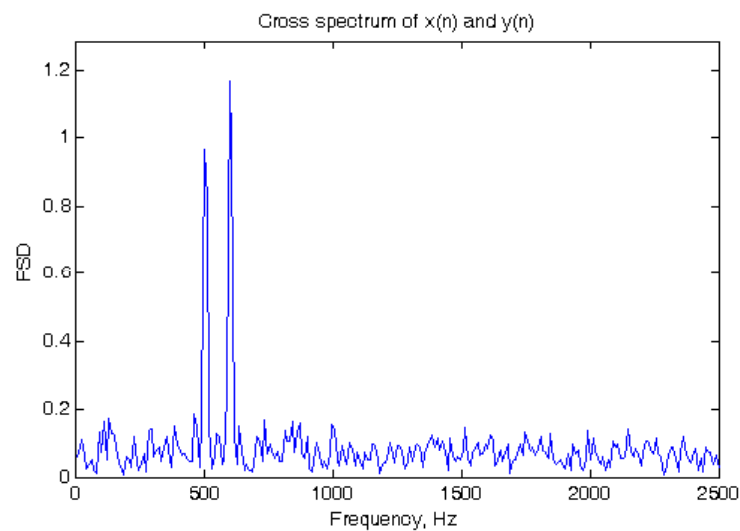
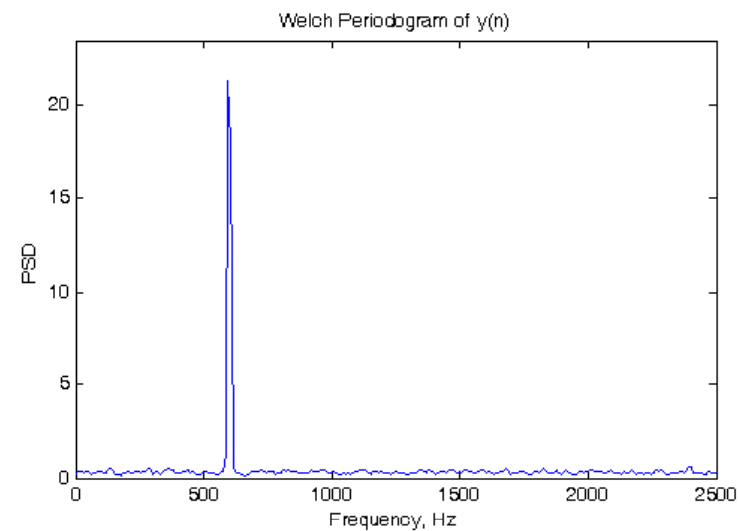
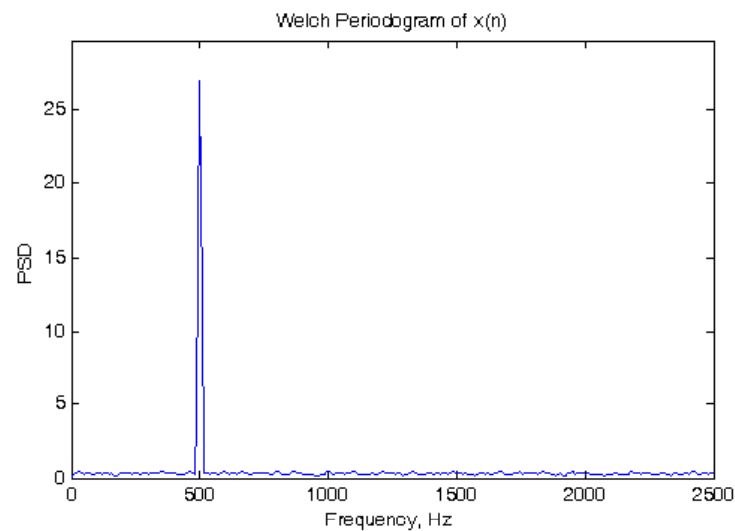
% Coherence

```
[Cxy,f] = mscohere(x,y,hanning(NFFT),NFFT/2,NFFT);  
f=f/pi*fs/2;  
plot(f,Cxy)
```

Cross spectrum and coherence of two sinusoids with identical frequency

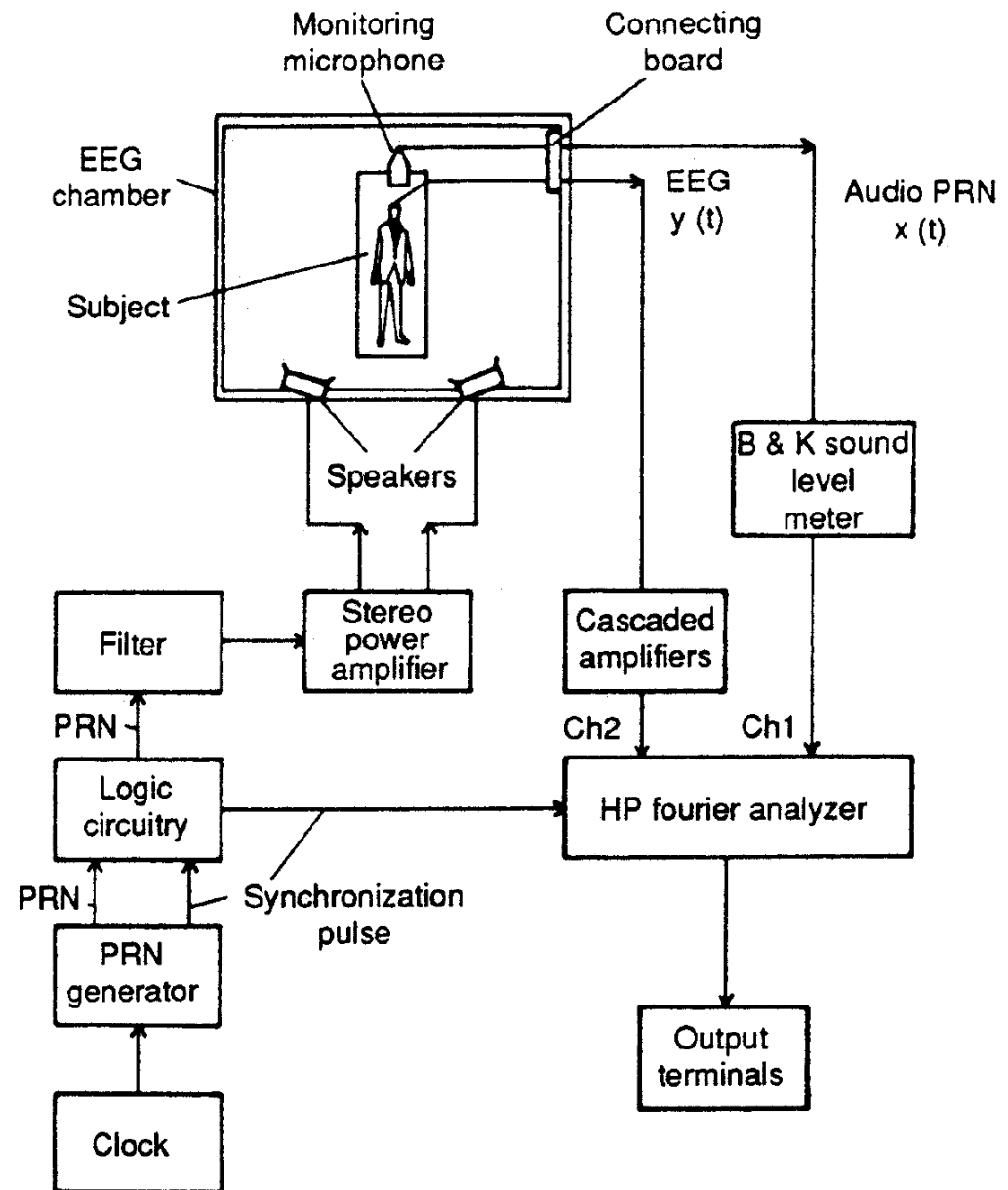


Cross spectrum and coherence of two sinusoids with different frequency

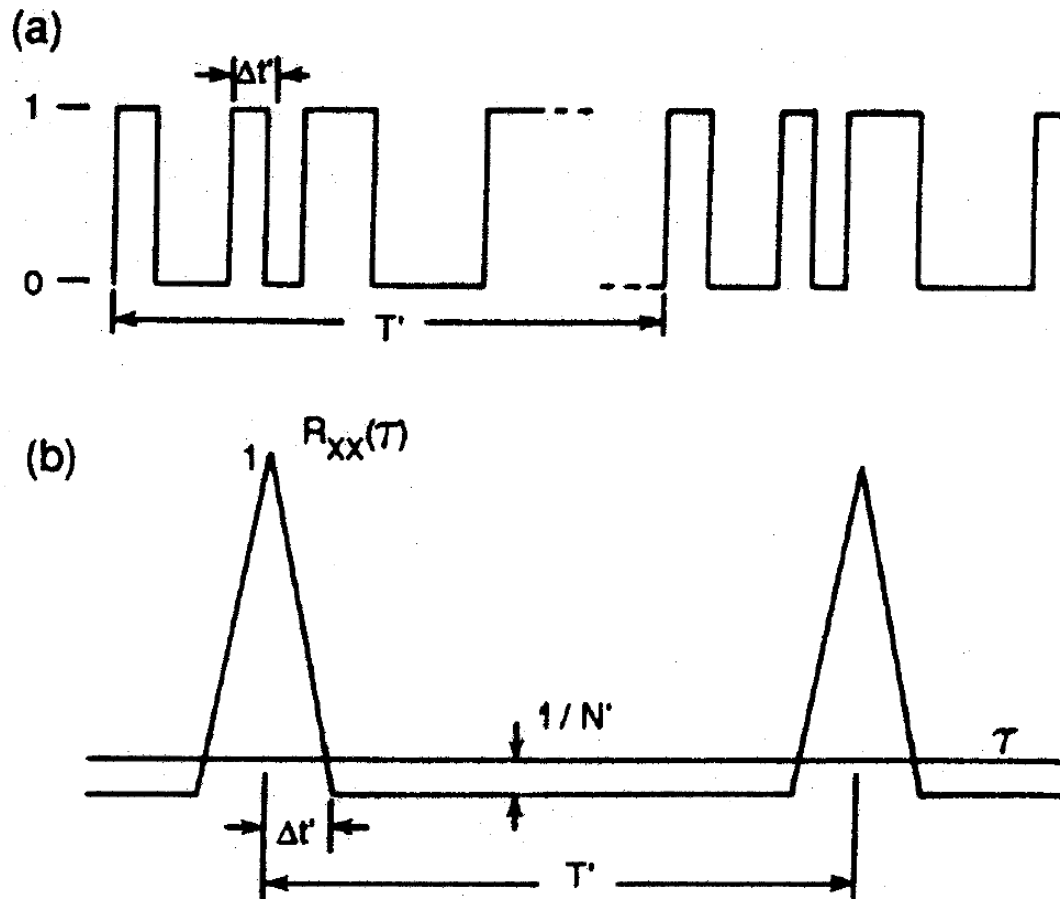


Analysis of auditory evoked potentials

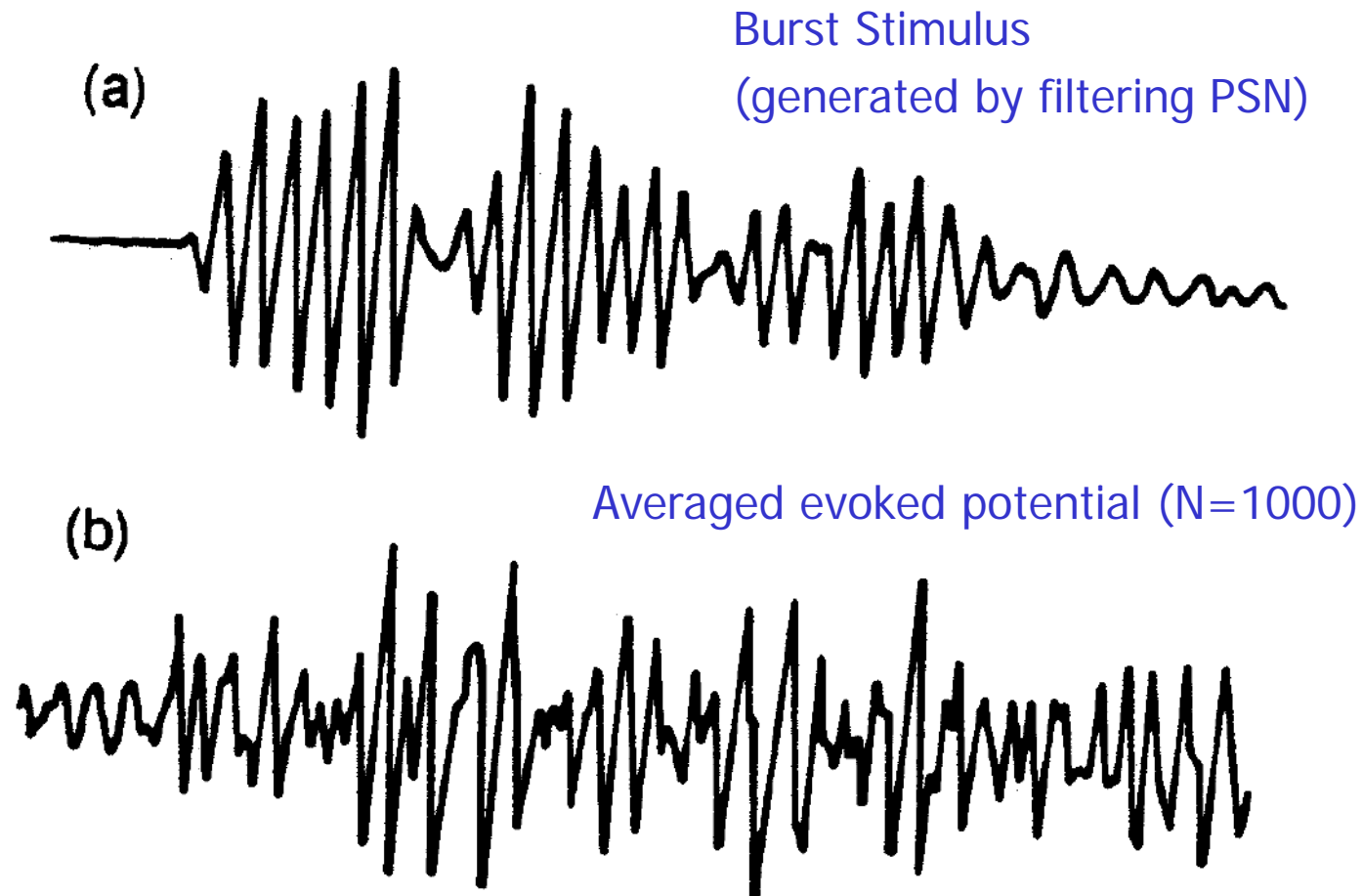
SM. Reddy, RL. Kirlin, Spectral analysis of auditory evoked potentials with pseudorandom random noise excitation, IEEE Trans. Biomed Eng, 26(8):479-487, 1979.



Pseudorandom noise (PSN)



Narrow-band burst stimulus at 600 Hz in one normal subject

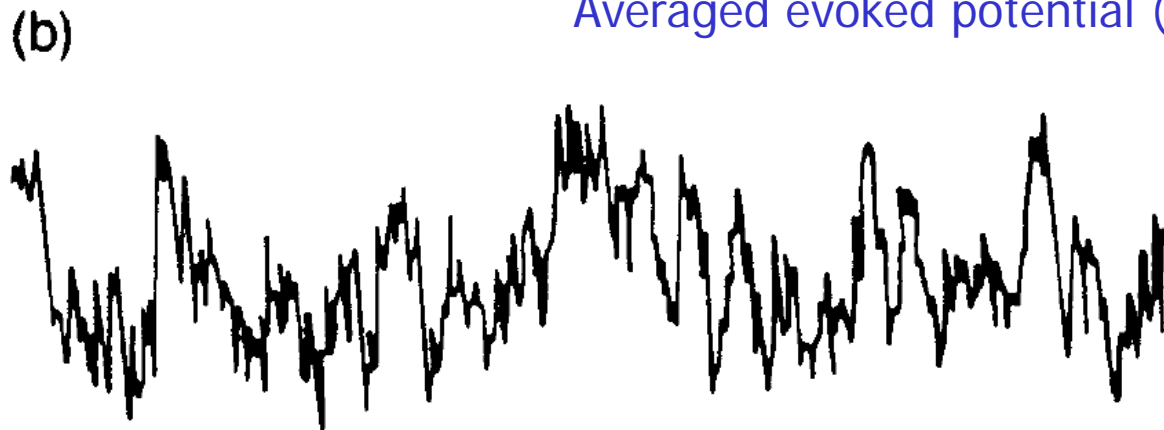


Narrow-band burst stimulus at 600 Hz in one patient with hearing loss

Burst Stimulus

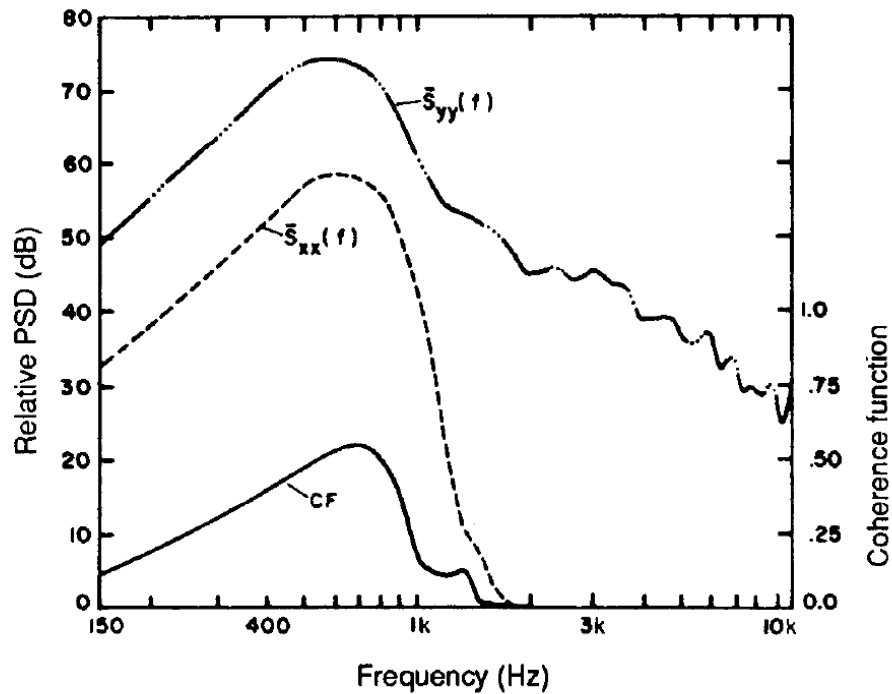


Averaged evoked potential (N=1000)

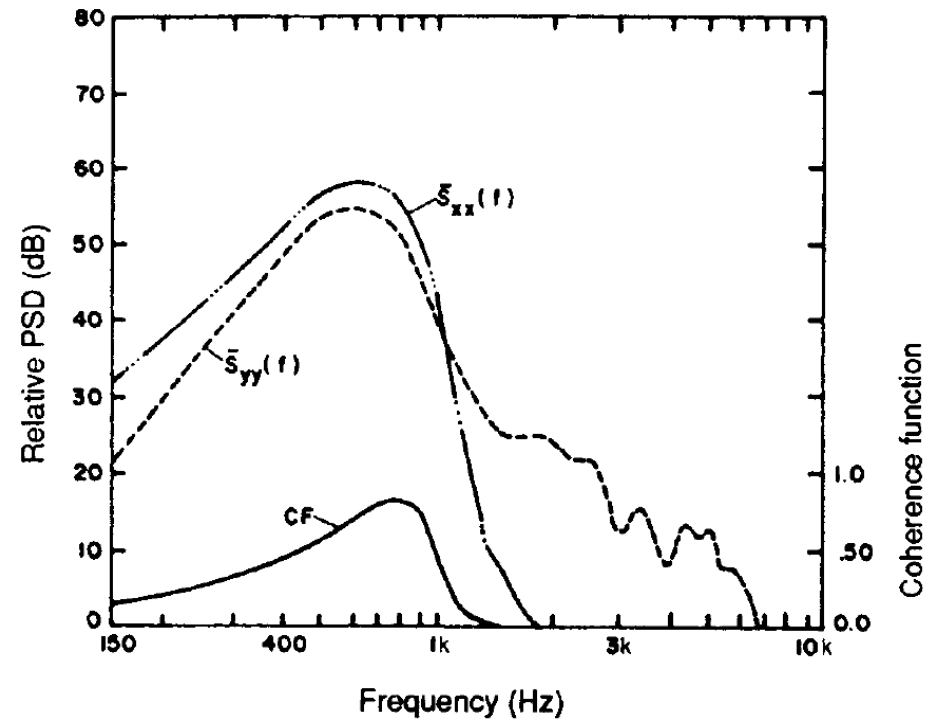


Power spectral density and coherence using 600-Hz stimulus

Normal Subject

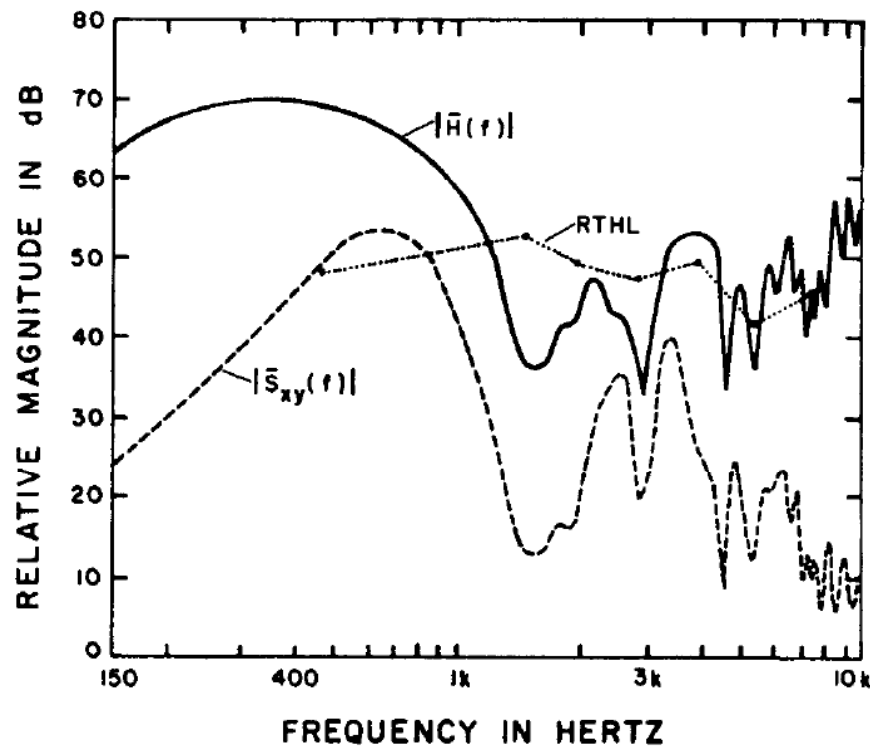


Patient with Hearing Loss

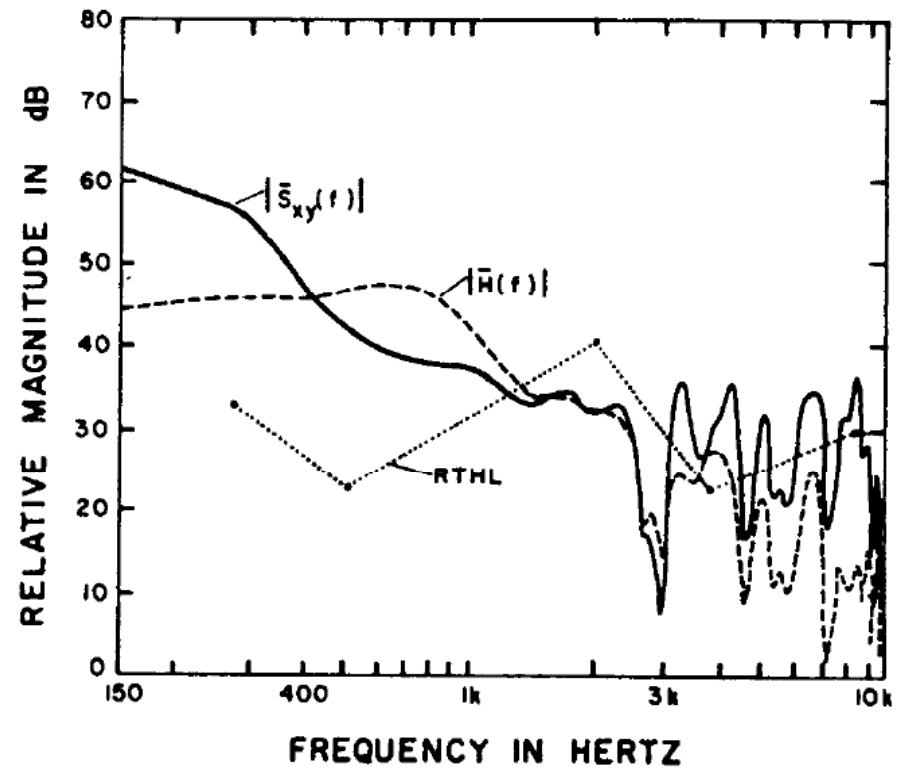


Cross spectral density and transfer function

Normal subject



Patient with hearing loss



Reference

- J.H. McClellan, R.W. Schafer, M.A. Yoder, Signal Processing First, Prentice Hall, 2003.
- R. Rangayyan, Biomedical Signal Analysis, John Wiley & Sons, 2002.
- J. Semmlow, Circuits, Signals, and Systems for Bioengineers: A MATLAB-Based Introduction, Academic Press, 2005.
- M.J. Roberts, Signals and Systems: Analysis of Signals Through Linear Systems, McGraw-Hill, 2003.