

# Lecture 7

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## Periodogram

A periodogram can estimate the Power Spectrum Density of a signal from a finite set of  $N$  observations given by:

$$I(\omega) = \frac{1}{N} \left| \sum_{n=0}^{N-1} x[n] e^{-j\omega n} \right|^2 \quad (14.34)$$

The periodogram can estimate of a zero-mean wide-sense stationary process (with absolute summable autocorrelation sequence) is a continuous ("smooth") function of  $\omega$  defined by:

$$S_{xx}(\omega) = \sum_{\ell=-\infty}^{\infty} r_{xx}(\ell) e^{-j\omega \ell}$$

## Modified Periodogram

$$\tilde{I}(\omega) \triangleq \frac{1}{N} \left| \sum_{n=0}^{N-1} w[n] x[n] e^{-j\omega n} \right|^2 \quad (14.52)$$

## Functions

```
function r=acrs(x, L)
    % Computes the ACRS r[m] for 0 <= m <= L
    % r=acrs(x-mean(x),L) yields the ACVS
    N=length(x);
```

```

x1=zeros(N+L-1,1);
x2=x1;
x1(1:N,1)=x;
for m=1:L
    x2=zeros(N+L-1,1);
    x2(m:N+m-1,1)=x;
    r(m)=x1'*x2;
end
r=r(:)/N;
end

```

```

function I=psdper(x, K)
    % Compute periodogram I(w) using the FFT.
    % K-point FFT >= N
    N=length(x);
    X=fft(x,K);
    I=X.*conj(X)/N;
    I(1)=I(2); % Avoid DC bias
    I=I(:);
end

```

```

function r=acrsfft(x, L)
    % Compute the autocorrelation sequence using the FFT.
    % r=acrsfft(x-mean(x),L) yields the ACVS
    N=length(x);
    Q=nextpow2(N+L);
    X=fft(x,2^Q);
    r0=real(ifft(X.*conj(X)));
    r=r0(1:L)/N;
end

```

```

function I=psdmodper(x, K)
    % Compute the modified periodogram PSD estimate.
    % K-point FFT >= N
    N=length(x);
    w=hann(N); % choose window
    w=w/(norm(w)/sqrt(N)); % sum w^2[k]=N
    X=fft(x(:).*w(:),K);
    I=X.*conj(X)/N;
    I(1)=I(2); % Avoid DC bias
    I=I(:);
end

```

```

function psdbt(x, L, K)
    % Compute the Blackman-Tukey PSD estimate.
    % Blackman-Tukey PSD estimator of S(2*pi*k/K)
    N=length(x);
    w=hann(N); % Data window
    w=w/(norm(w)/sqrt(N)); % sum w^2[k]=N
    x=x.*w; % Data windowing
    r=acrsfft(x,L);

```

```

wc=parzenwin(2*L-1); % Lag window
rw=r.*wc(L:2*L-1); % Lag windowing
g=zeros(K,1);
g(1:L)=rw;
g(K-L+2:K)=flipud(rw(2:L));
G=fft(g,K); % f even => F real
S=2*real(G(1:K/2)); S(1)=S(2);
end

function S=psdwelch(x, L, K)
% Compute the Welch PSD estimate.
% Welch PSD estimator of  $S(2\pi k/K)$ 
M=fix((length(x)-L/2)/(L/2)) % 50% overlap
time=(1:L)';
I=zeros(K,1);
w=hanning(L); % Choose window
w=w/(norm(w)/sqrt(L)); % sum  $w^2[k]=L$ 
for m=1:M
% xw=w.*detrend(x(time)); % detrending
xw=w.*x(time); % data windowing
X=fft(xw,K);
I=I+X.*conj(X);
time=time+L/2;
end
I=I/(M*L); S=2*I(1:K/2); S(1)=S(2);
end

```