



Conventional Power Spectral Estimation

Power Spectral Estimation

Periodogram is an estimate of the power spectrum

$$I_N(\omega) = \sum_{m=-(N-1)}^{N-1} c_{xx}(m) e^{-j\omega m} \quad c_{xx}(m) = \frac{1}{N} \phi(m)$$

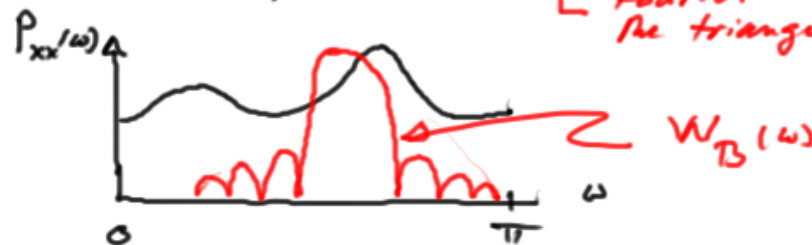
$$= \frac{1}{N} |X(e^{j\omega})|^2 \quad \text{where: } X(e^{j\omega}) = \sum_{n=0}^{N-1} x(n) e^{-j\omega n}$$

$$\frac{\text{Bias}}{E[I_N(\omega)]} = \sum_{m=-(N-1)}^{N-1} \left(\frac{N-|m|}{N} \right) \phi_{xx}(m) e^{-j\omega m}$$

← triangular window

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} P_{xx}(\theta) W_B(e^{j(\omega-\theta)}) d\theta$$

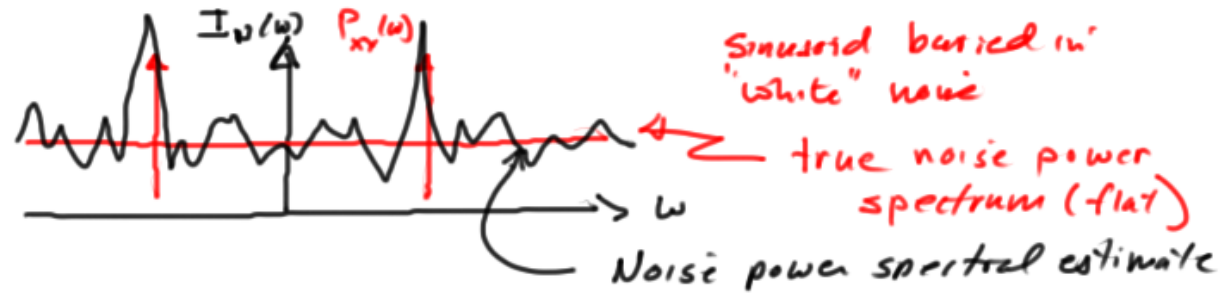
← Fourier transform of the triangular window





Conventional Power Spectral Estimation

Variance



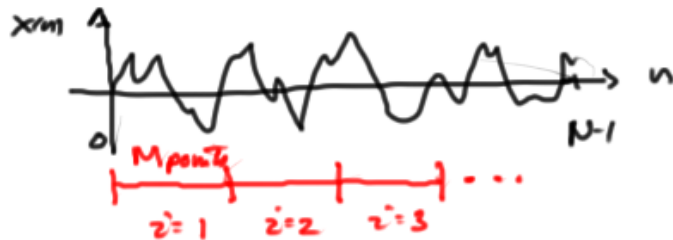
$$\text{var} [I_N(\omega)] = P_{xx}^2(\omega) \left\{ 1 + \left(\frac{\sin \omega N}{N \sin \omega} \right)^2 \right\}$$

Thus, variance of periodogram does not get smaller as N increases



Conventional Power Spectral Estimation

Bartlett's Procedure of Averaging Periodograms



$$B_{xx}(\omega) = \frac{1}{K} \sum_{z=1}^K I_M^{(z)}(\omega)$$

$$E[B_{xx}(\omega)] = \frac{1}{K} \sum_{z=1}^K E[I_M^{(z)}(\omega)] = E[I_M^{(z)}(\omega)]$$

Convolution of $P_{xx}(\omega)$ with the Fourier transform of triangular window $\left(\frac{M-|m|}{M}\right)$

Variance

$$\text{var}[B_{xx}(\omega)] = \frac{1}{K} \text{var}[I_M^{(z)}(\omega)]$$

Variance reduction resulting from averaging $\rightarrow \frac{1}{K} P_{xx}^e(\omega)$



Conventional Power Spectral Estimation

Welch's Method of Averaging Modified Periodograms

$$B_{xx}^w(\omega) = \frac{1}{K} \sum_{i=1}^K J_M^{(i)}(\omega)$$

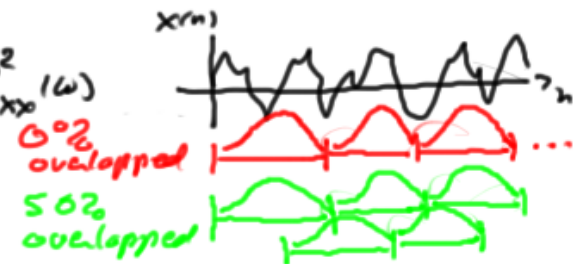
$$\text{where: } J_M^{(i)}(\omega) = \frac{1}{Mu} \left| \sum_{n=0}^{M-1} w(n) x^{(i)}(n) e^{-j\omega n} \right|^2$$

$$\text{and } u = \frac{1}{M} \sum_{n=0}^{M-1} w^2(n)$$

$$E[B_{xx}^w(\omega)] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P_{xx}(\theta) W(e^{j(\omega-\theta)}) d\theta$$

$$W(e^{j\omega}) = \frac{1}{Mu} \left| \sum_{n=0}^{M-1} w(n) e^{-j\omega n} \right|^2$$

$$\text{var}[B_{xx}^w(\omega)] \approx \frac{1}{K} P_{xx}^2(\omega)$$

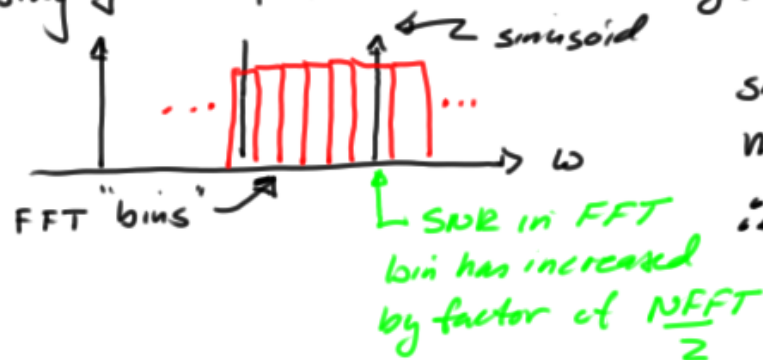




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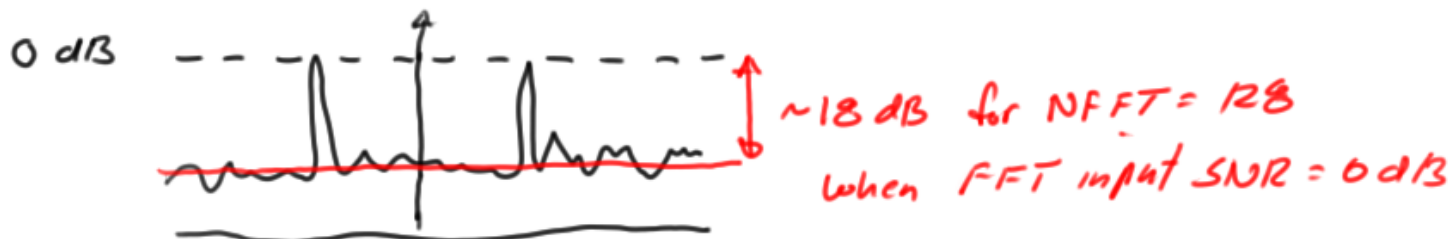
HW #4 - Spectral Averaging

Processing gain of FFT (dB) $\approx 10 \log_{10} \frac{N_{FFT}}{2}$



sinusoid power split into 2 bins
noise power split into N bins
 \therefore SNR gain is $\frac{N_{FFT}}{2}$

$N_{FFT} = 128$	$\frac{N_{FFT}}{2} = 64 = 2^6$	SNR gain = 18 dB
256	$128 = 2^7$	21 dB
512	$256 = 2^8$	24 dB
1024	$512 = 2^9$	27 dB





Conventional Power Spectral Estimation

HW #4 - Spectral Averaging

C. Incoherent Average = $\frac{1}{K} \sum_{i=1}^K |X_{(k)}^{(i)}|^2$



↑
M = 128 point FFTs

... for 0% overlap

$X_{(k)}^{(1)} \quad X_{(k)}^{(2)}$

D. Coherent Average = $\frac{1}{K} \sum_{i=1}^K X_{(k)}^{(i)}$

to compute power then take $| \cdot |^2$ of result

= $\sum_{n=0}^{M-1} \left\{ \frac{1}{K} \sum_{i=1}^K x_{(n)}^{(i)} \right\} e^{-j \left(\frac{2\pi}{M} \right) nk}$

↑
time domain

frequency domain

bin centered sinusoid

non-bin centered sinusoid