

# SNR Calculation and Spectral Estimation [S&T Appendix A]

or, How *not* to make a mess of an FFT

- 0 Make sure the input is located in an FFT bin**
- 1 Window the data!**  
A Hann window works well.
- 2 Compute the FFT**
- 3 SNR = power in signal bins / power in noise bins**
- 4 If you want to make a spectral plot**
  - i. Apply sine-wave scaling
  - ii. State the noise bandwidth (NBW)
  - iii. Smooth the FFT

# FT and DFT (1)

- Fourier Transform:

$$x(t) \leftrightarrow X(\omega)$$

- If  $x(t)$  is sampled

$$x(nT) \leftrightarrow \sum_{n=-\infty}^{\infty} x(nT) \cdot e^{-j\omega nT}$$

- Estimation of spectrum: DFT / FFT

$$X(f_k) = \sum_{n=0}^{N-1} x(nT) \cdot e^{-j2\pi n T f_k} = x(nT) * h_k(n)$$

where  $f_k = k / (NT)$ ,  $k=0, 1, 2, \dots, N-1$

$$h_k(n) = \begin{cases} e^{j2\pi kn/N}, & 0 \leq n \leq N \\ 0 & \text{otherwise} \end{cases}$$

## FT and DFT (2)

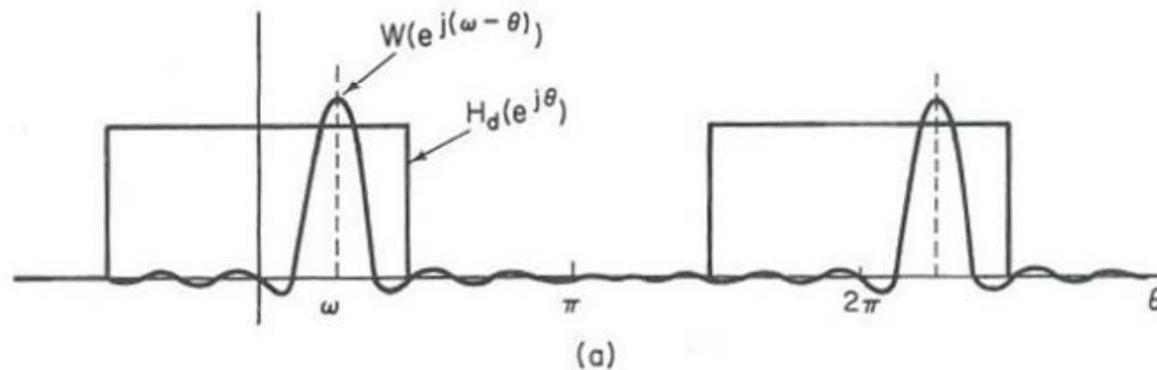
- Generally, in Fourier Transformation, the rule is

*Sampled  $\leftrightarrow$  Periodic*

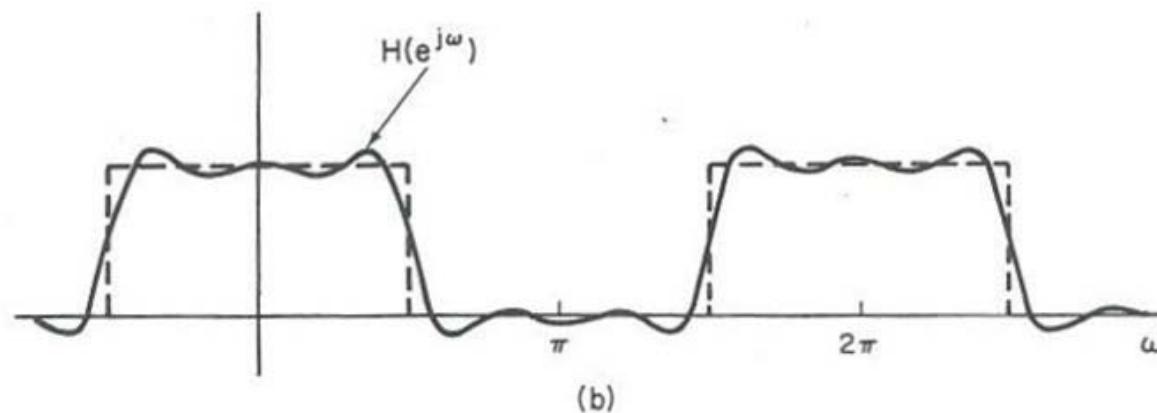
- If  $x(nT)$  is not periodic with period  $NT$ , the DFT calculates the spectrum of a discontinuous signal -- bad estimate!

# FT and DFT (3)

- Another problem: convolution introduces noise folding in windowed spectrum:



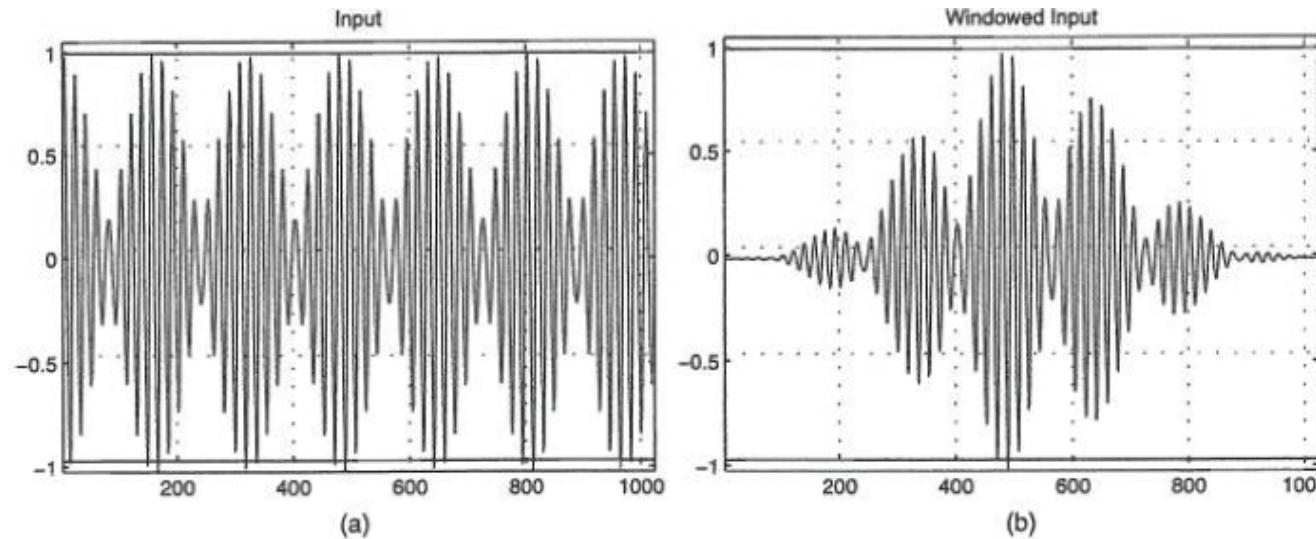
(a)



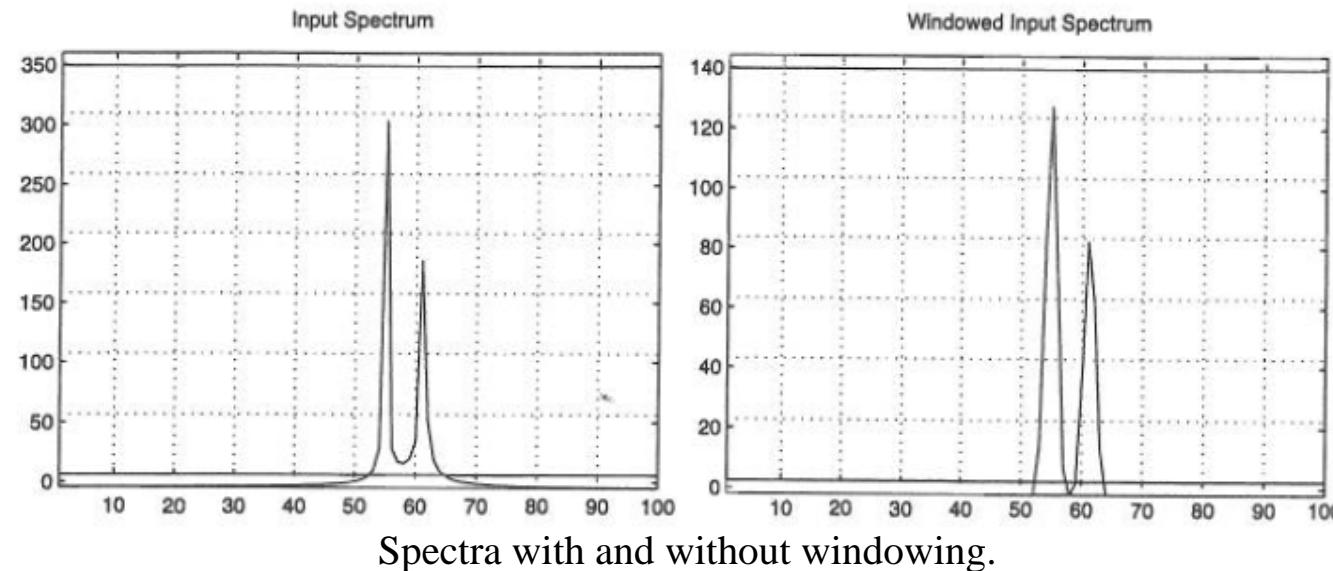
(b)

- (a) Convolution process implied by truncation of the ideal impulse response. (b) Typical approximation resulting from windowing the ideal impulse response.

# Example: two-tone signal



Time domain input of the example before (a) and after windowing (b).



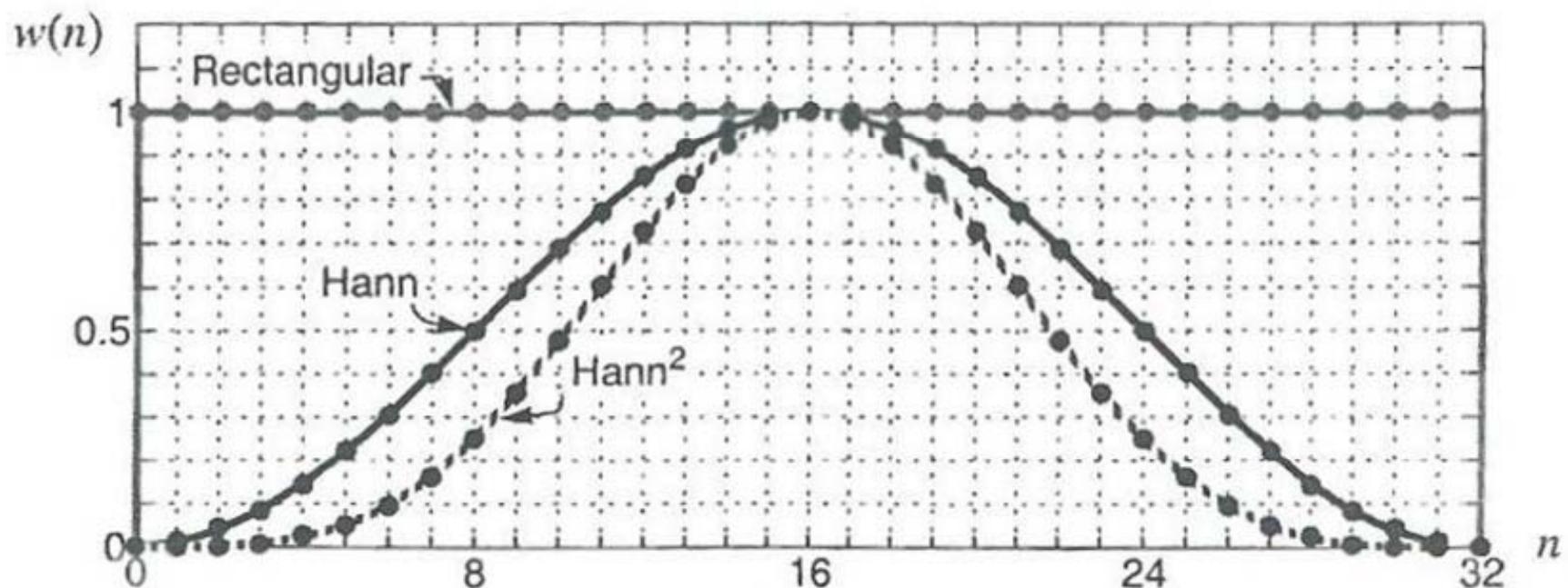
Spectra with and without windowing.

# Windowing

- General window ( $T=1$ ):

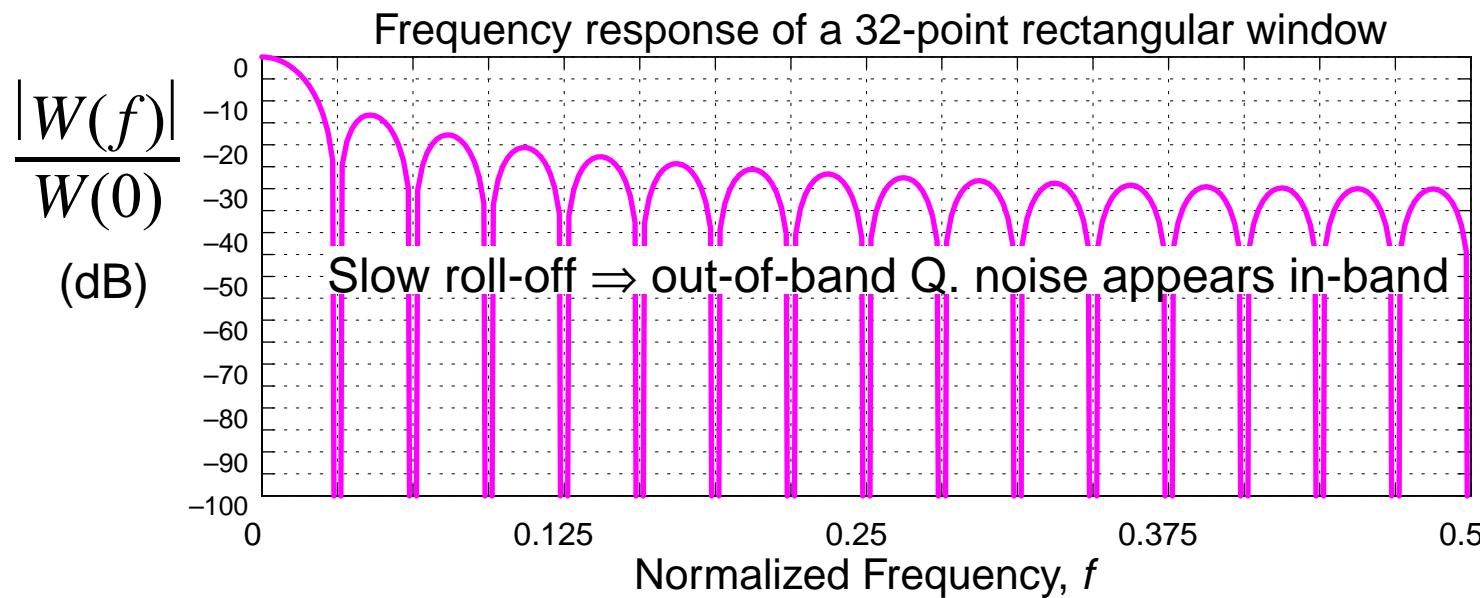
$$W(f) = \sum_{n=0}^{N-1} w(n) \cdot e^{-j2\pi fn}$$

- Common windows:



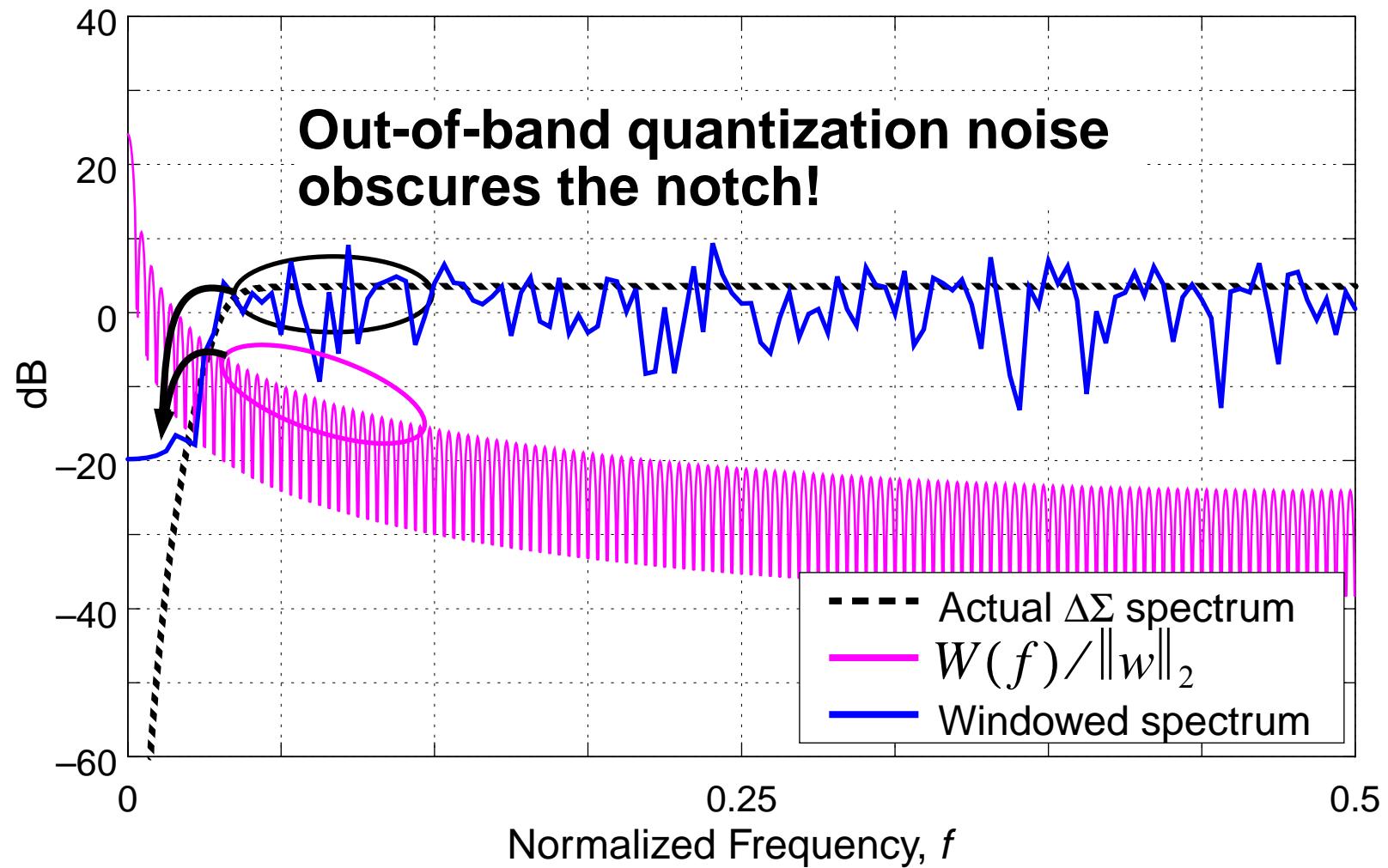
# Windowing

- **$\Delta\Sigma$  data is (usually) not periodic**  
Just because the input repeats does not mean that the output does too!
- **A finite-length data record = an infinite record multiplied by a *rectangular window*:  $w(n) = 1, 0 \leq n < N$**   
Windowing is unavoidable.
- **“Multiplication in time is convolution in frequency”**



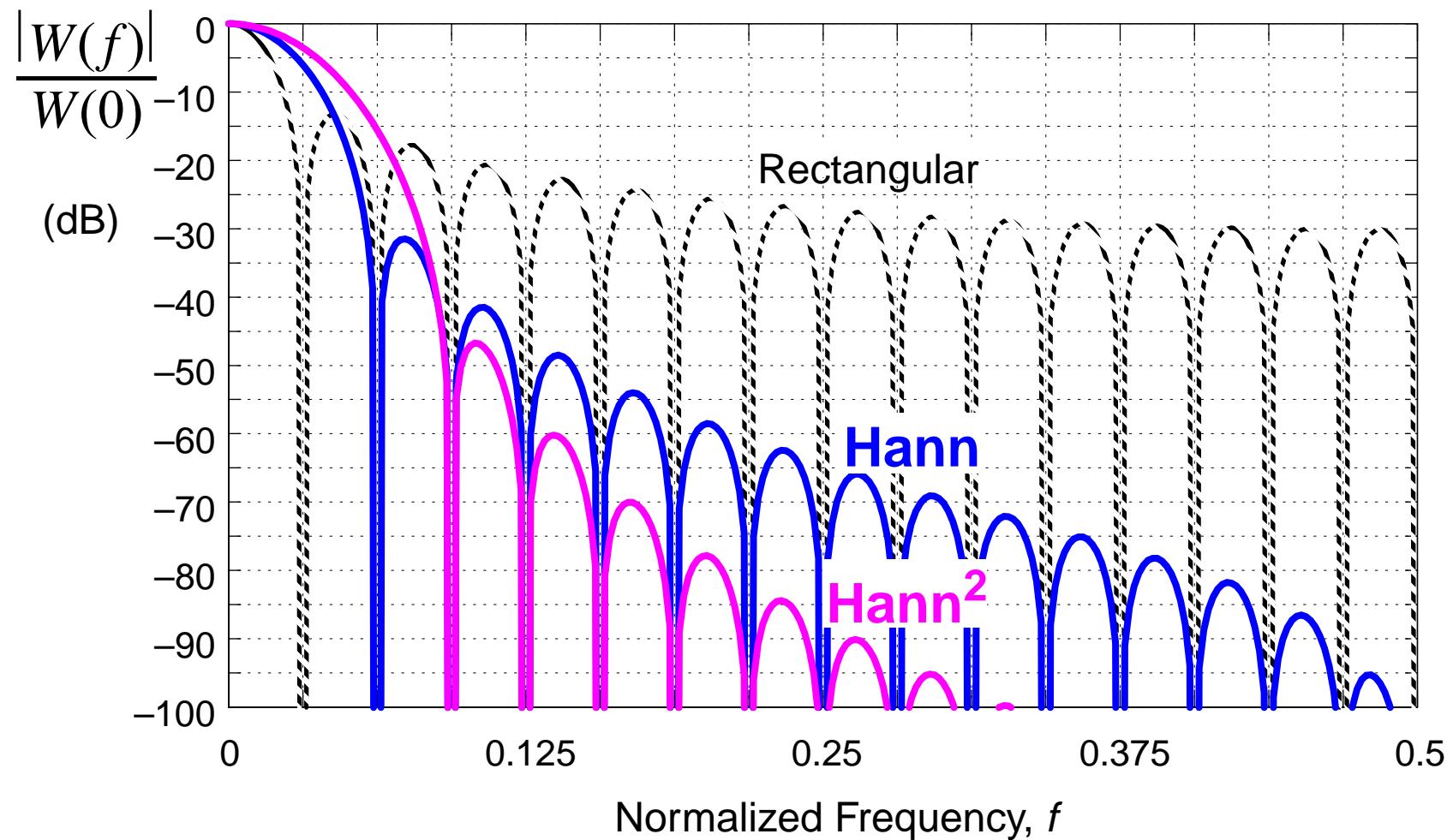
# Example Spectral Disaster

Rectangular window,  $N = 256$



# Window Comparison

$N = 16$



# Window Properties

Window	Rectangular	Hann <sup>†</sup>	Hann <sup>2</sup>
$w(n)$ , $n = 0, 1, \dots, N - 1$ ( $w(n) = 0$ otherwise)	1	$\frac{1 - \cos \frac{2\pi n}{N}}{2}$	$\left( \frac{1 - \cos \frac{2\pi n}{N}}{2} \right)^2$
Number of non-zero FFT bins	1	3	5
$\ w\ _2^2 = \sum w(n)^2$	$N$	$3N/8$	$35N/128$
$W(0) = \sum w(n)$	$N$	$N/2$	$3N/8$
$NBW = \frac{\ w\ _2^2}{W(0)^2}$	$1/N$	$1.5/N$	$35/18N$

<sup>†</sup>. MATLAB's hanning function causes spectral leakage of tones located in FFT bins unless you add the optional argument "periodic". Use ΔΣ Toolbox function ds\_hann.

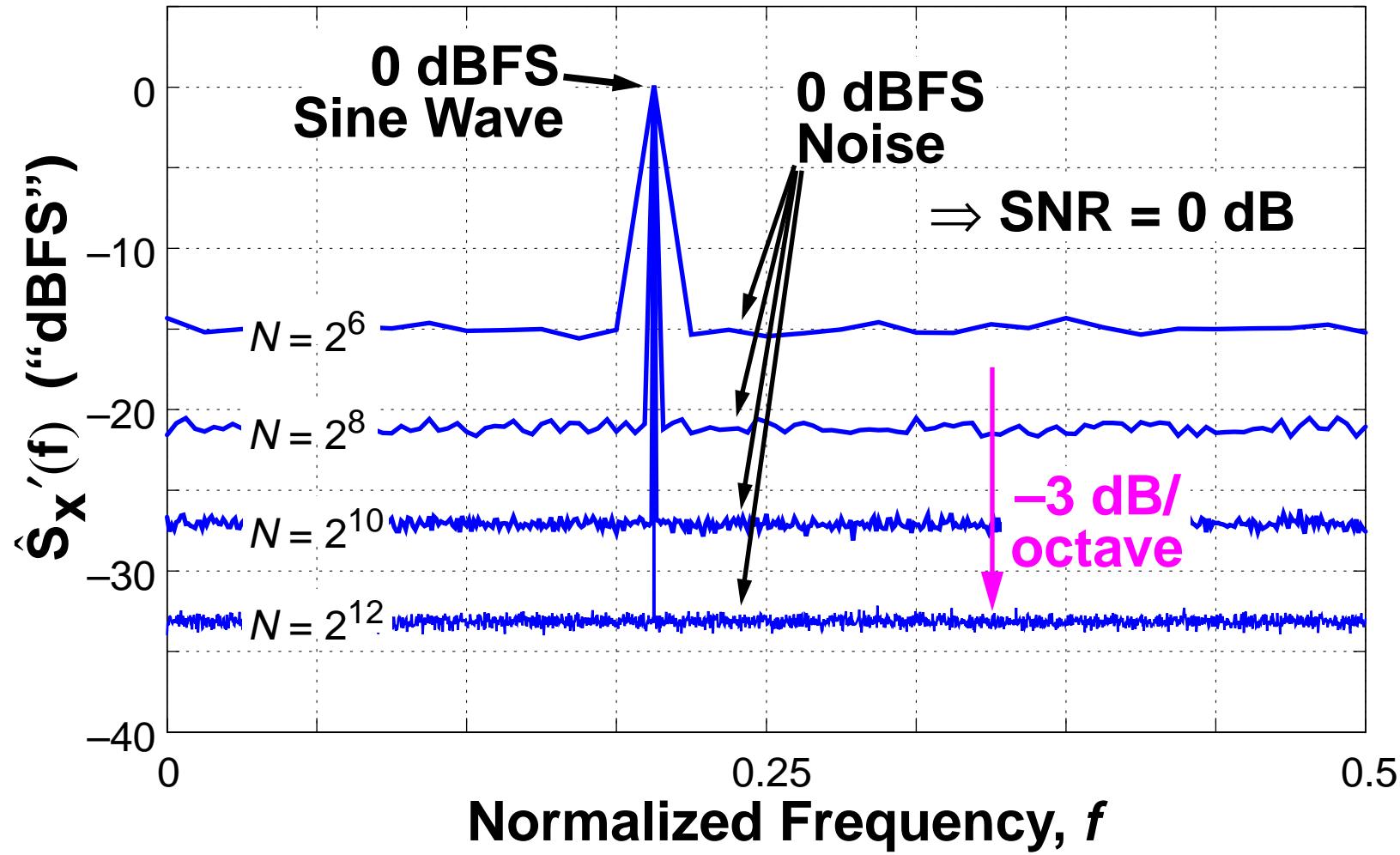
# Window Length, $N$

- Need to have enough in-band noise bins to
  - 1 Make the number of signal bins a small fraction of the total number of in-band bins  
 $<20\%$  signal bins  $\Rightarrow >15$  in-band bins  $\Rightarrow N > 30 \cdot OSR$
  - 2 Make the SNR repeatable
    - $N = 30 \cdot OSR$  yields std. dev.  $\sim 1.4$  dB.
    - $N = 64 \cdot OSR$  yields std. dev.  $\sim 1.0$  dB.
    - $N = 256 \cdot OSR$  yields std. dev.  $\sim 0.5$  dB.
- $N = 64 \cdot OSR$  is recommended

This is all you need to know to do SNR calculations.  
If you want to make spectral plots, you need to know more...

# Spectrum of a Sine Wave plus Noise

Various  $N$



# Scaling and Noise Bandwidth

- FFT scaled such that a full-scale (FS) sine wave ( $A = FS/2$ ) yields a 0-dB spectral peak:

$$\hat{S}_x'(f) = \frac{\left| \sum_{n=0}^{N-1} w(n) \cdot x(n) \cdot e^{-j2\pi fn} \right|^2}{(FS/4)W(0)}$$

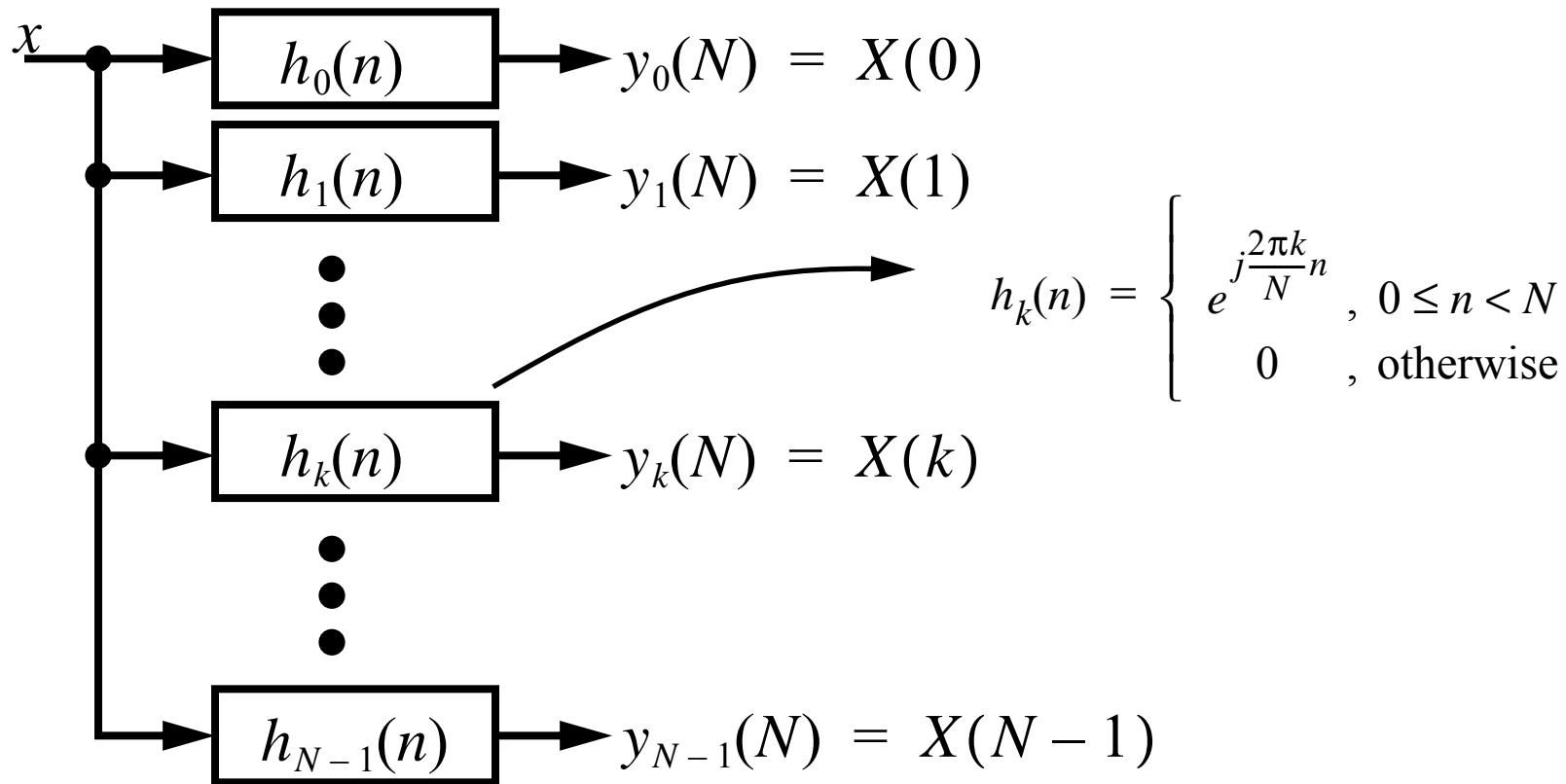
← **|FFT|<sup>2</sup>**  
← **sine-wave scale factor**

- “Noise Floor” depends on  $N$  (!)  
 A sine-wave-scaled FFT is fine for showing *spectra*,  
 but is ill-suited for displaying *spectral densities*.
- Vertical axis is really “dBFS/NBW,” where NBW is the bandwidth over which the noise power has been integrated

Think of the spectrum as representing the amount of power in a frequency band whose width is NBW.  
 $NBW = k/N$ , where  $k$  depends on the window type.

# An FFT is like a Filter Bank

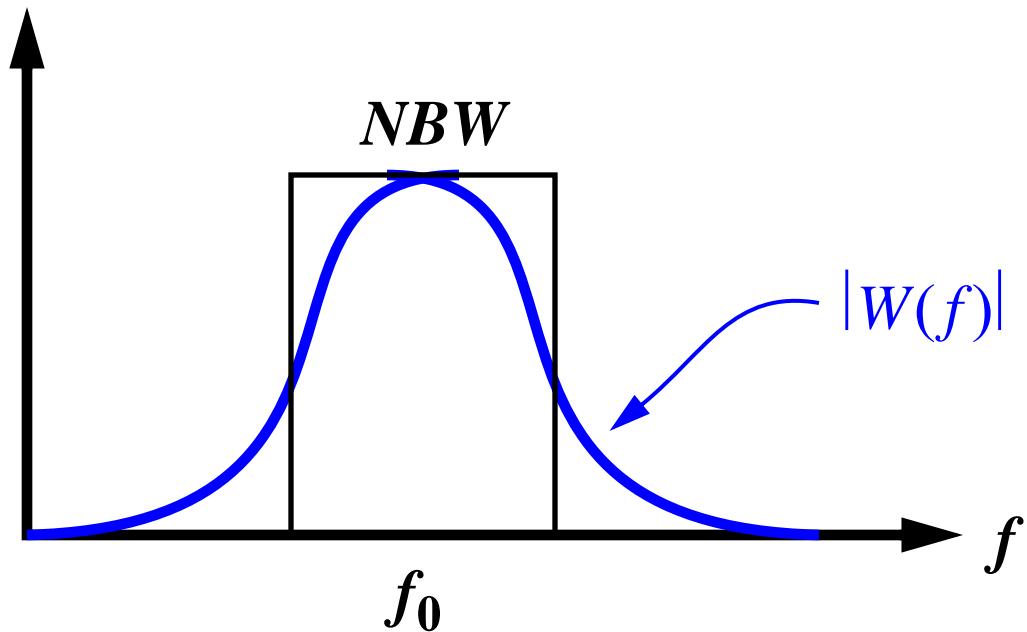
- The FFT can be interpreted as taking 1 sample from the outputs of  $N$  complex FIR filters:



- NBW is the effective bandwidth of these filters

# Noise Bandwidth

- For a filter with frequency response  $W(f)$



$$NBW = \frac{\int |W(f)|^2 df}{W(f_0)^2}$$

# Noise Bandwidth of a Rectangular FFT

$$h_k(n) = \exp\left(j\frac{2\pi k}{N}n\right)$$

$$W_k(f) = \sum_{n=0}^{N-1} h_k(n) \exp(-j2\pi fn)$$

$$f_0 = \frac{k}{N}, \quad W_k(f_0) = \sum_{n=0}^{N-1} 1 = N$$

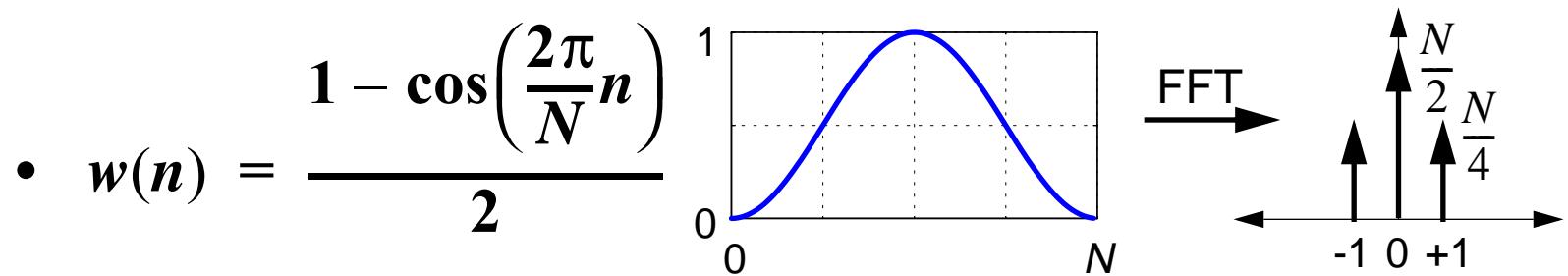
$$\int |W_k(f)|^2 = \sum |w_k(n)|^2 = N \quad [\text{Parseval}]$$

$$\therefore NBW = \frac{\int |W_k(f)|^2 df}{W_k(f_0)^2} = \frac{N}{N^2} = \frac{1}{N}$$

- NBW is the same for each FFT bin “filter”**

# Noise Bandwidth of a Hann-Windowed FFT

- Use the filter associated with FFT bin 0

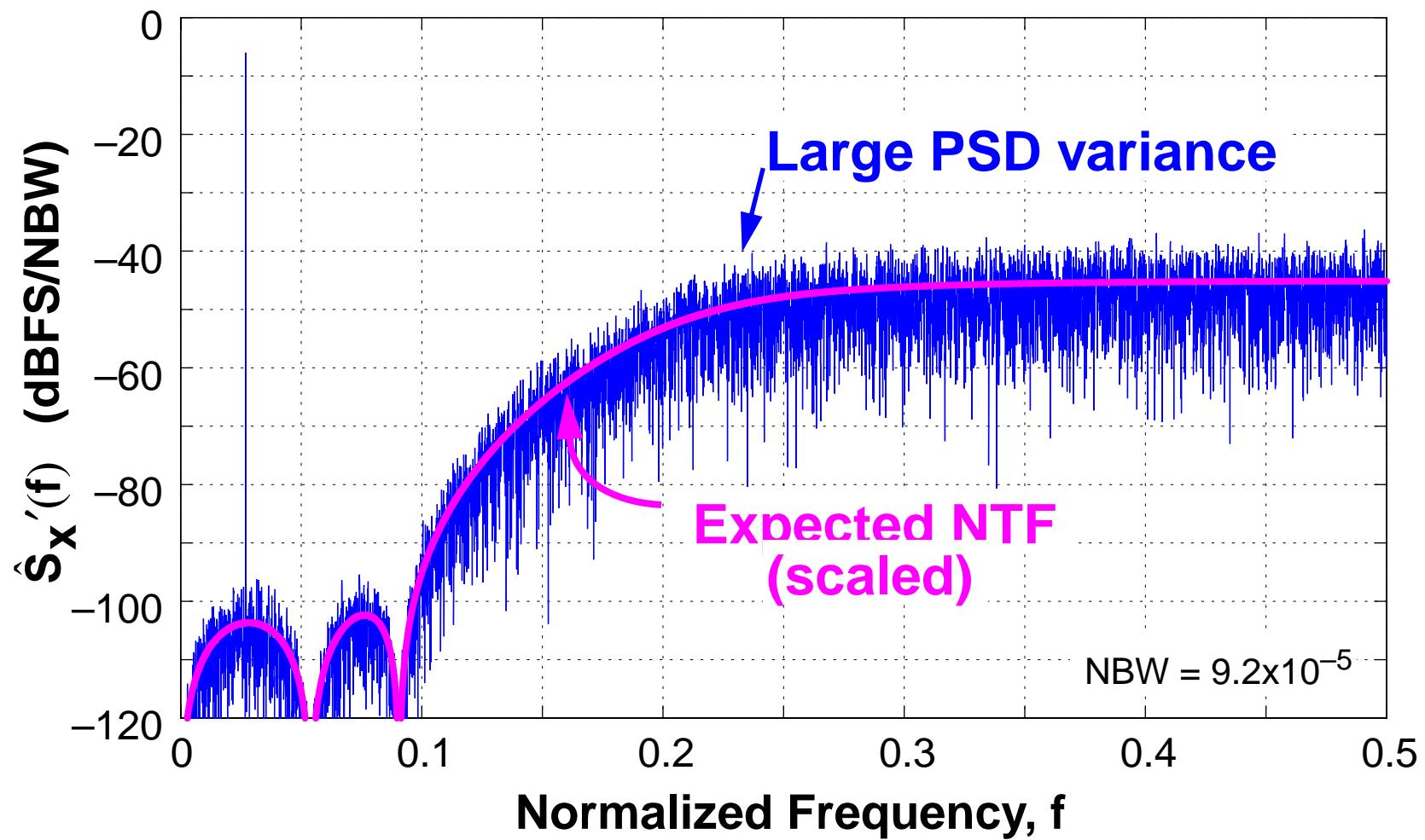


$$\Rightarrow \|w\|_2^2 = \frac{3N}{8}, W(0) = \frac{N}{2}$$

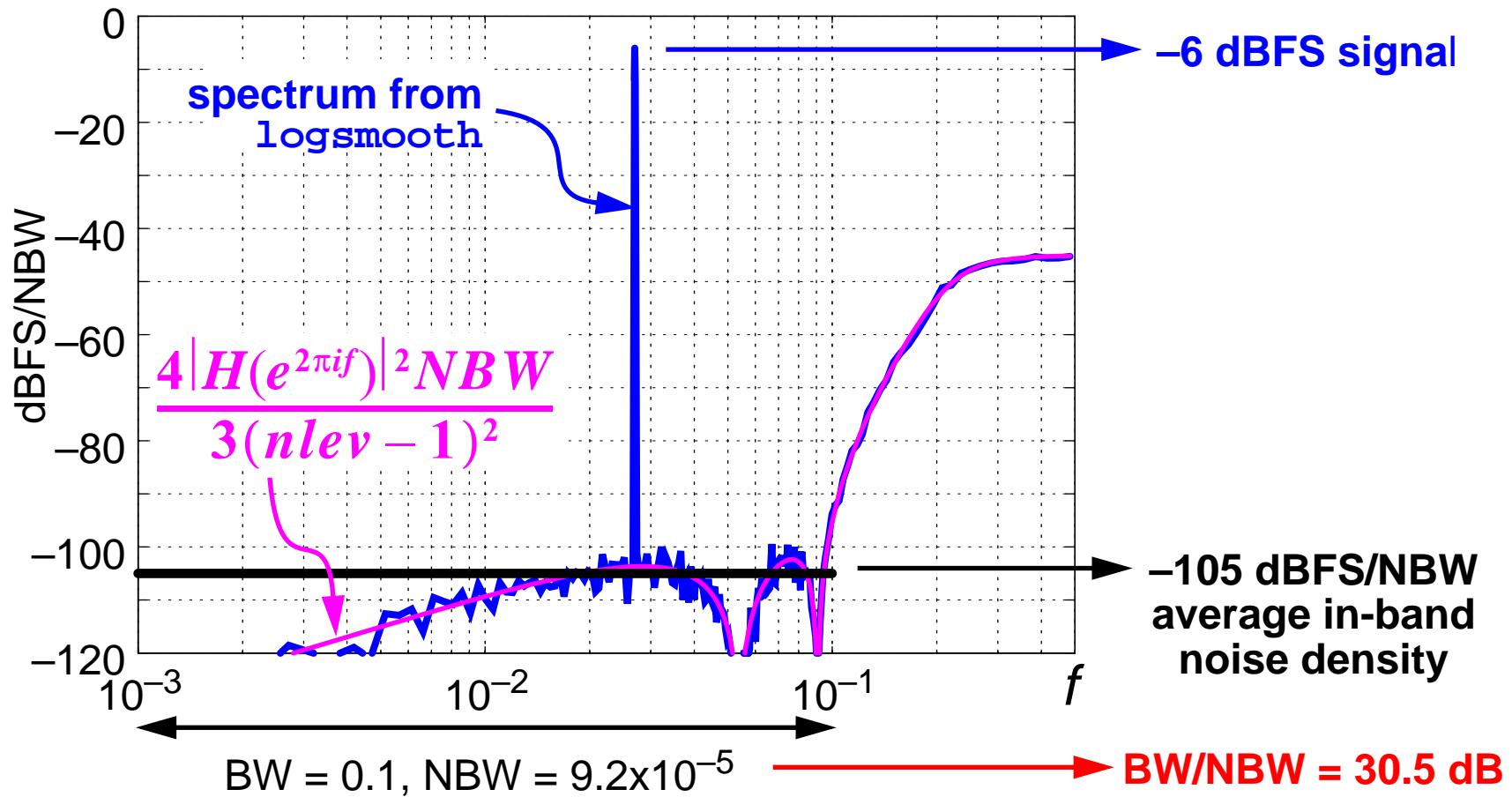
$$\Rightarrow NBW = \frac{\|w\|_2^2}{W(0)^2} = \frac{1.5}{N}$$

# Example $\Delta\Sigma$ Spectrum

5<sup>th</sup>-order NTF, OSR=5, Hann window, no averaging



# Smoothed Spectrum and SNR Calculation



- Quantization Noise Power =  $-105 + 30.5 = -74.5$  dBFS  
 $\Rightarrow$  SQNR =  $-6 - (-74.5) = 68.5$  dB

# Manual SQNR Prediction

- The noise term is  $HE$ .
- The rms value of  $H$  in the band of interest,  $\sigma_H$ , can be evaluated using `rmsGain`.
- Since  $\Delta = 2$  for all quantizers,  $\sigma_e^2 = \frac{\Delta^2}{12} = \frac{1}{3}$ .
- The in-band noise power is therefore  $\frac{\sigma_H^2 \sigma_e^2}{\text{OSR}} = \frac{\sigma_H^2}{3\text{OSR}}$ .
- The signal power is impossible to predict using the linear model, but is usually around  $-3$  dBFS.  
This corresponds to a power of  $(nlev - 1)^2 / 4$ .
- $\therefore \text{SQNR}_{\text{peak}} \approx 10 \log_{10} \left( \frac{3(\text{OSR})(nlev - 1)^2}{4\sigma_H^2} \right) \text{ dB.}$