

# ECE1371 Advanced Analog Circuits

## Lecture 1

# INTRODUCTION TO DELTA-SIGMA ADCS

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## Course Goals

- **Deepen understanding of CMOS analog circuit design through a top-down study of a modern analog system— a delta-sigma ADC**
- **Develop circuit insight through brief peeks at some nifty little circuits**

The circuit world is filled with many little gems that every competent designer ought to know.

# Logistics

- **Format:**  
Meet Mondays 1:00-3:00 PM (not Feb 16, Apr 6)  
12 2-hr lectures plus proj. presentation
- **Grading:**  
30% homework  
40% project  
30% exam
- **References:**  
Schreier & Temes, “Understanding  $\Delta\Sigma$  ...”  
Chan-Carusone, Johns & Martin, “Analog IC ...”  
Razavi, “Design of Analog CMOS ICs”
- **Lecture Plan:**

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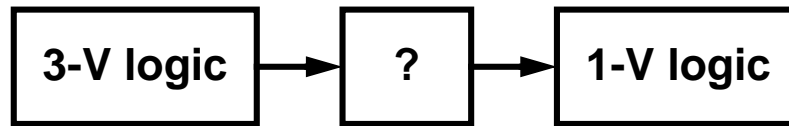
Date	Lecture (M 13:00-15:00)		Ref	Homework	
2015-01-05	RS	1	MOD1 & MOD2	ST 2, 3, A	1: Matlab MOD1&2
2015-01-12	RS	2	MODN + $\Delta\Sigma$ Toolbox	ST 4, B	2: $\Delta\Sigma$ Toolbox
2015-01-19	RS	3	Example Design: Part 1	ST 9.1, CCJM 14	3: Sw.-level MOD2
2015-01-26	RS	4	Example Design: Part 2	CCJM 18	
2015-02-02	TC	5	SC Circuits	R 12, CCJM 14	4: SC Integrator
2015-02-09	TC	6	Amplifier Design		
2015-02-16	Reading Week– No Lecture				
2015-02-23	TC	7	Amplifier Design		5: SC Int w/ Amp
2015-03-02	RS	8	Comparator & Flash ADC	CCJM 10	Project
2015-03-09	TC	9	Noise in SC Circuits	ST C	
2015-03-16	RS	10	Advanced $\Delta\Sigma$	ST 6.6, 9.4	
2015-03-23	TC	11	Matching & MM-Shaping	ST 6.3-6.5, +	
2015-03-30	TC	12	Pipeline and SAR ADCs	CCJM 15, 17	
2015-04-06	Exam		Proj. Report Due Friday April 10		
2015-04-13	Project Presentation				

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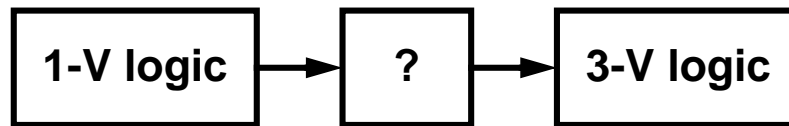
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# NLCOTD: Level Translator

VDD1 > VDD2, e.g.



- VDD1 < VDD2, e.g.



- Constraints: **CMOS**  
1-V and 3-V devices  
no static current

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## Highlights (i.e. What you will learn today)

- 1 **MOD1: 1<sup>st</sup>-order  $\Delta\Sigma$  modulator**  
Structure and theory of operation
- 2 **Inherent linearity of binary modulators**
- 3 **Inherent anti-aliasing of continuous-time modulators**
- 4 **MOD2: 2<sup>nd</sup>-order  $\Delta\Sigma$  modulator**
- 5 **Good FFT practice**

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# 0. Background

## (Stuff you already know)

- The SQNR\* of an ideal  $n$ -bit ADC with a full-scale sine-wave input is  $(6.02n + 1.76)$  dB  
“6 dB = 1 bit.”
- The PSD at the output of a linear system is the product of the input's PSD and the squared magnitude of the system's frequency response

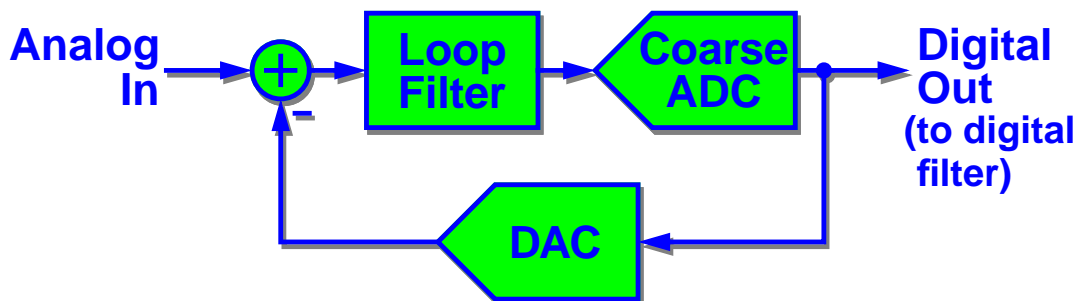
i.e.  $X \rightarrow \boxed{H(z)} \rightarrow Y$   $S_{yy}(f) = |H(e^{j2\pi f})|^2 \cdot S_{xx}(f)$

- The power in any frequency band is the integral of the PSD over that band

\*. SQNR = Signal-to-Quantization-Noise Ratio

## 1. What is $\Delta\Sigma$ ?

- $\Delta\Sigma$  is NOT a fraternity
- Simplified  $\Delta\Sigma$  ADC structure:



- Key features: coarse quantization, filtering, feedback and oversampling  
Quantization is often *quite* coarse (1 bit!), but the effective resolution can still be as high as 22 bits.

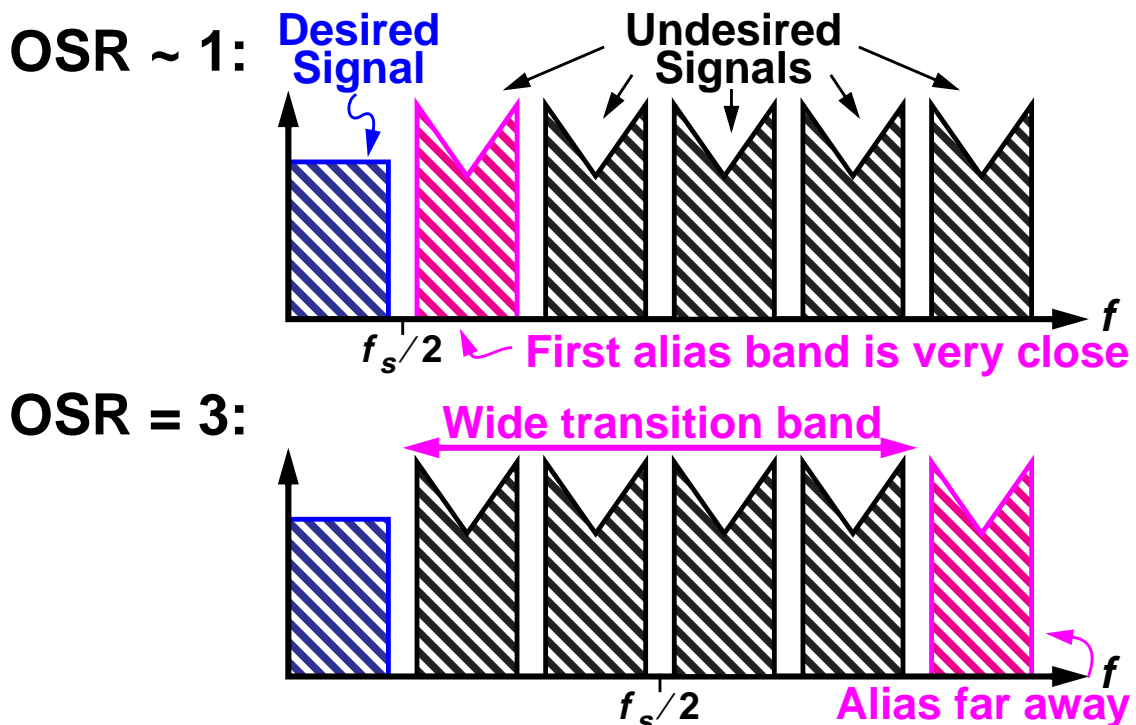
# What is Oversampling?

- **Oversampling is sampling faster than required by the Nyquist criterion**  
For a lowpass signal containing energy in the frequency range  $(0, f_B)$ , the minimum sample rate required for perfect reconstruction is  $f_s = 2f_B$
- **The *oversampling ratio* is  $OSR \equiv f_s / (2f_B)$**
- **For a regular ADC,  $OSR \sim 2 - 3$**   
To make the anti-alias filter (AAF) feasible
- **For a  $\Delta\Sigma$  ADC,  $OSR \sim 30$**   
To get adequate quantization noise suppression.  
Signals between  $f_B$  and  $\sim f_s$  are removed digitally.

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## Oversampling Simplifies AAF

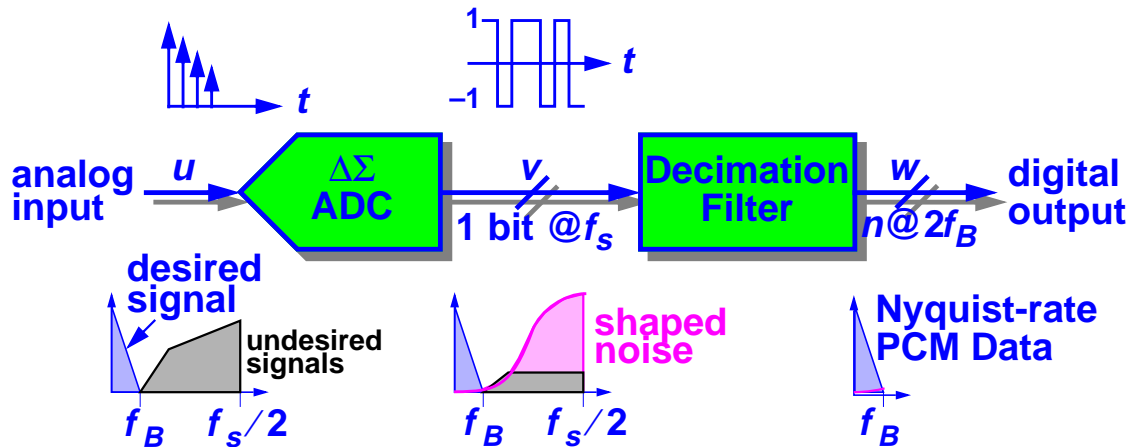


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# How Does A $\Delta\Sigma$ ADC Work?

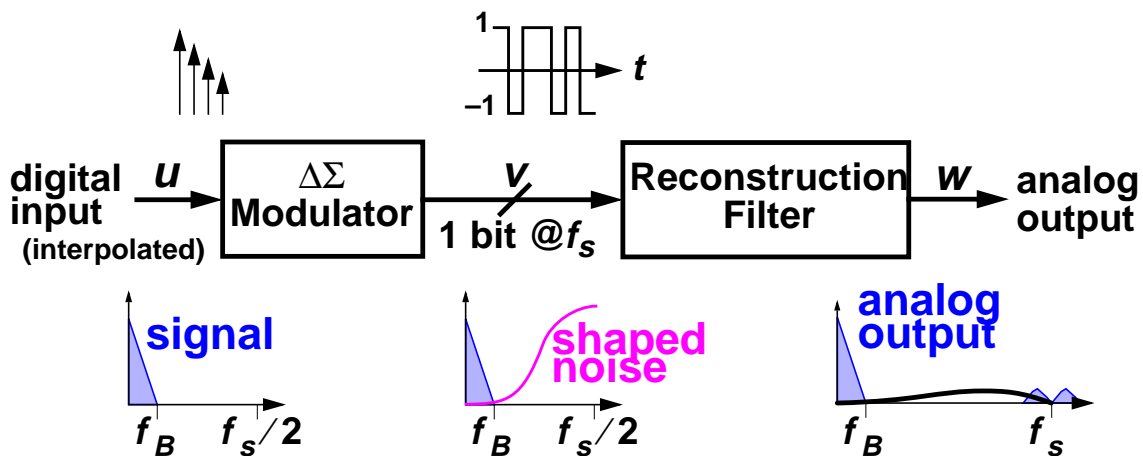
- Coarse quantization  $\Rightarrow$  lots of quantization error. So how can a  $\Delta\Sigma$  ADC achieve 22-bit resolution?
- A  $\Delta\Sigma$  ADC spectrally separates the quantization error from the signal through *noise-shaping*



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# A $\Delta\Sigma$ DAC System



- Mathematically similar to an ADC system  
 Except that now the modulator is digital and drives a low-resolution DAC, and that the out-of-band noise is handled by an analog reconstruction filter.

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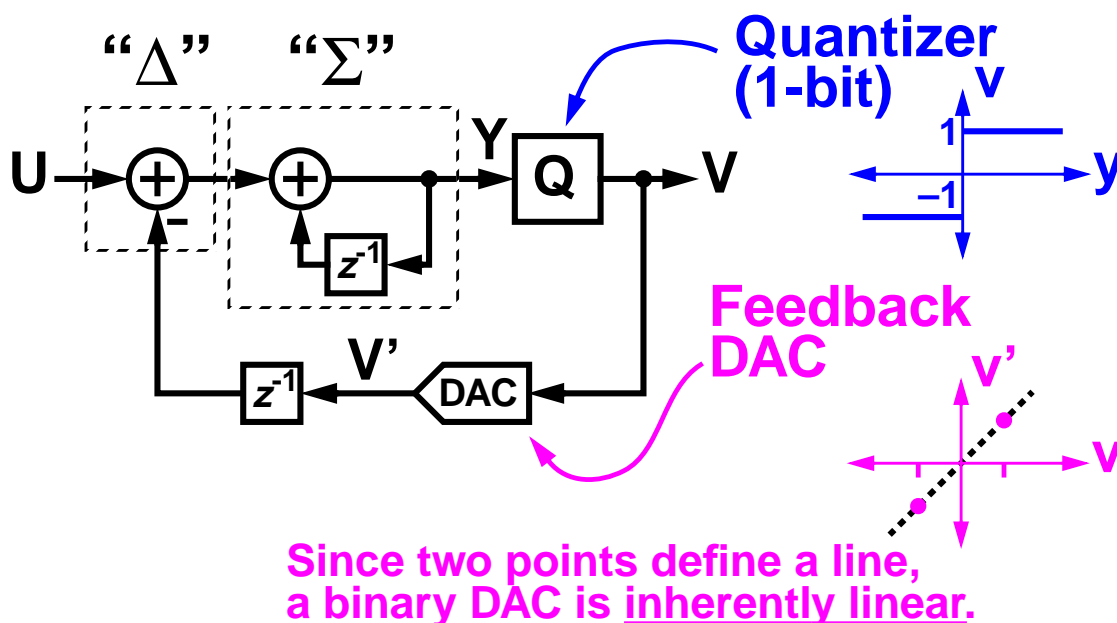
# Why Do It The $\Delta\Sigma$ Way?

- **ADC: Simplified Anti-Alias Filter**  
Since the input is oversampled, only very high frequencies alias to the passband.  
A simple RC section often suffices.  
If a continuous-time loop filter is used, the anti-alias filter can often be eliminated altogether.
- **DAC: Simplified Reconstruction Filter**  
The nearby images present in Nyquist-rate reconstruction can be removed digitally.
- + **Inherent Linearity**  
Simple structures can yield very high SNR.
- + **Robust Implementation**  
 $\Delta\Sigma$  tolerates sizable component errors.

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## 2. MOD1: 1<sup>st</sup>-Order $\Delta\Sigma$ Modulator [Ch. 2 of Schreier & Temes]

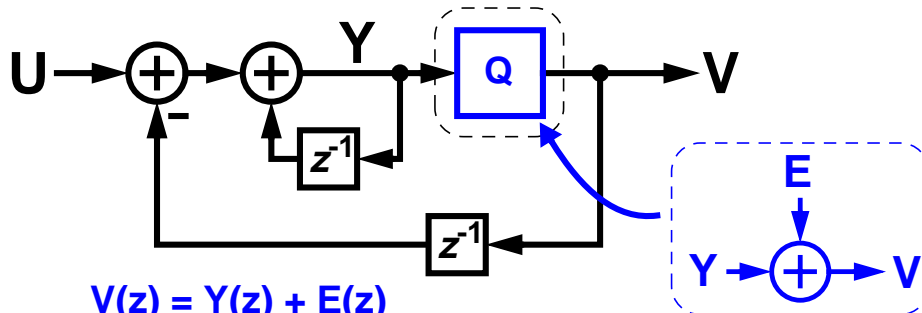


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# MOD1 Analysis

- Exact analysis is intractable for all but the simplest inputs, so treat the quantizer as an additive noise source:



$$V(z) = Y(z) + E(z)$$

$$Y(z) = (U(z) - z^{-1}V(z)) / (1 - z^{-1})$$

$$\Rightarrow (1 - z^{-1})V(z) = U(z) - z^{-1}V(z) + (1 - z^{-1})E(z)$$

$$V(z) = U(z) + (1 - z^{-1})E(z)$$

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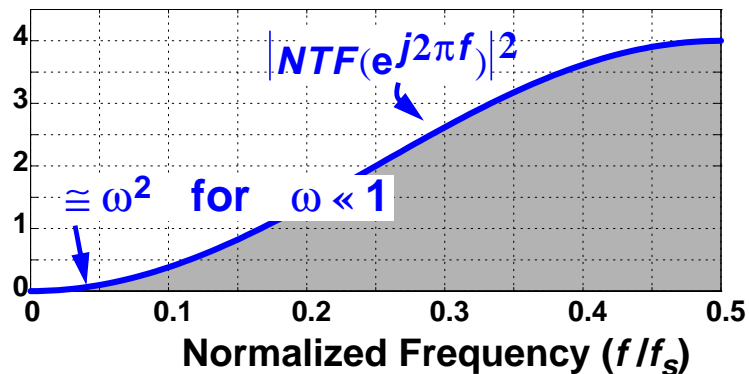
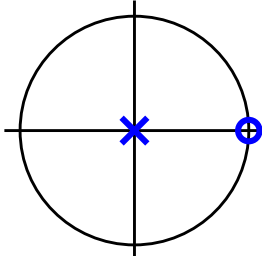
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## The Noise Transfer Function (NTF)

- In general,  $V(z) = \text{STF}(z) \cdot U(z) + \text{NTF}(z) \cdot E(z)$
- For MOD1,  $\text{NTF}(z) = 1 - z^{-1}$

The quantization noise has spectral shape!

Poles & zeros:



- The total noise power increases, but the noise power at low frequencies is reduced

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# In-band Quant. Noise Power

- Assume that  $e$  is white with power  $\sigma_e^2$   
i.e.  $S_{ee}(\omega) = \sigma_e^2/\pi$
- The in-band quantization noise power is

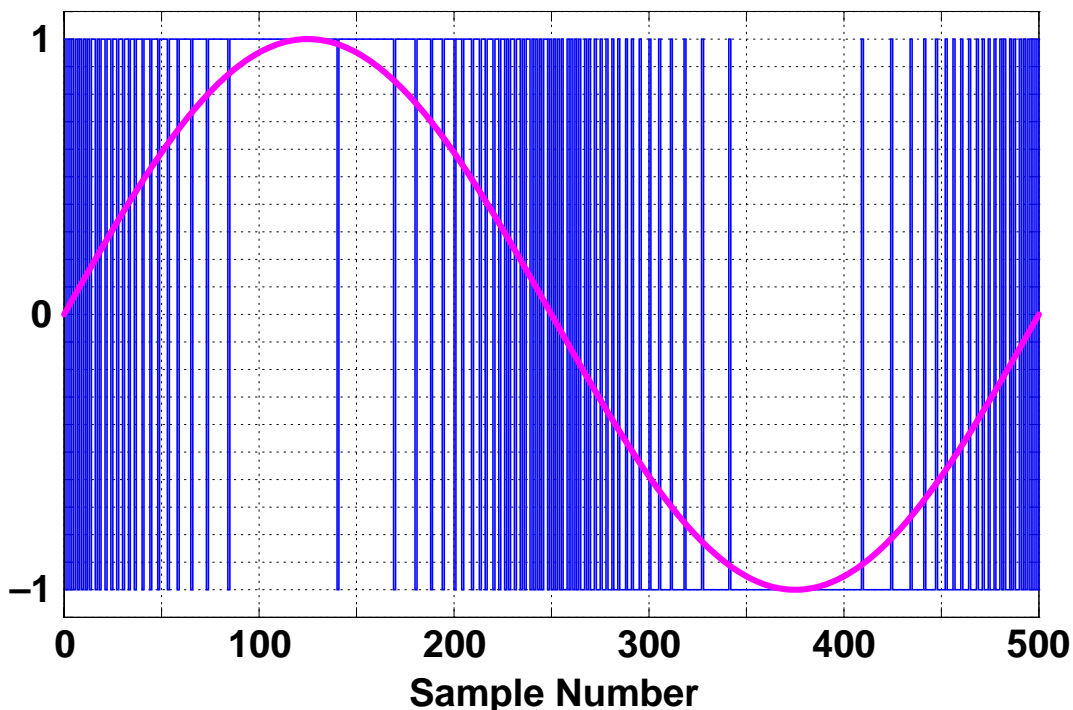
$$IQNP = \int_0^{\omega_B} |H(e^{j\omega})|^2 S_{ee}(\omega) d\omega \cong \frac{\sigma_e^2}{\pi} \int_0^{\omega_B} \omega^2 d\omega$$

- Since  $OSR \equiv \frac{\pi}{\omega_B}$ ,  $IQNP = \frac{\pi^2 \sigma_e^2}{3} (OSR)^{-3}$
- For MOD1, an octave increase in  $OSR$  increases SQNR by 9 dB  
“1.5-bit/octave SQNR-OSR trade-off.”

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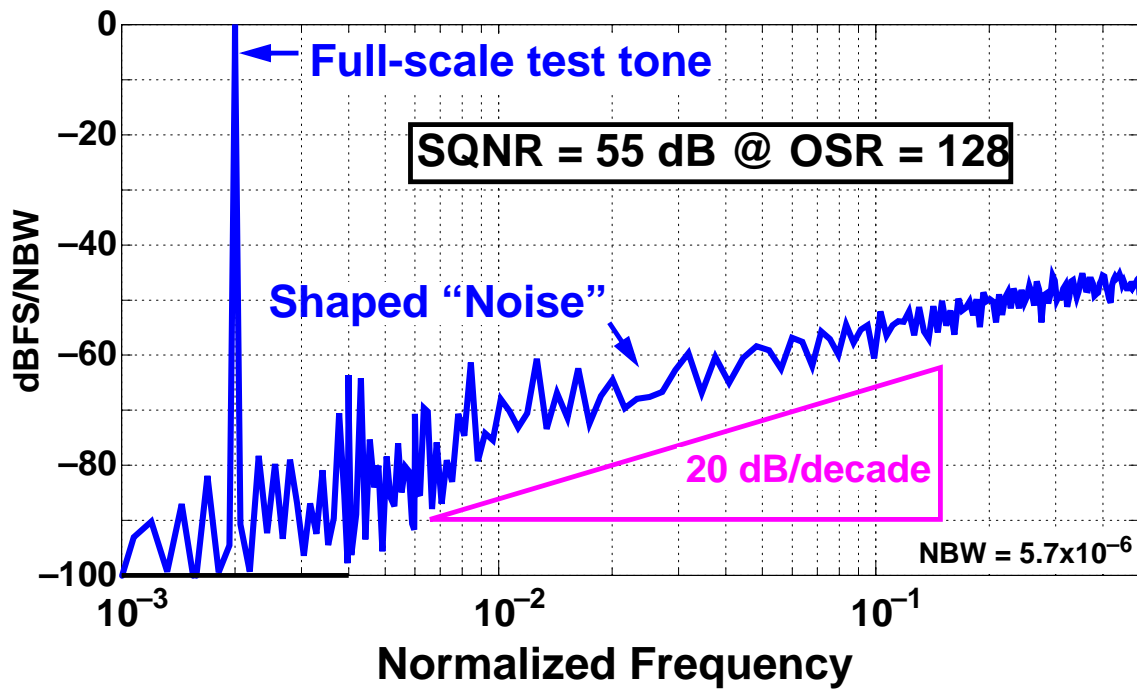
## A Simulation of MOD1— Time



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# A Simulation of MOD1— Freq.

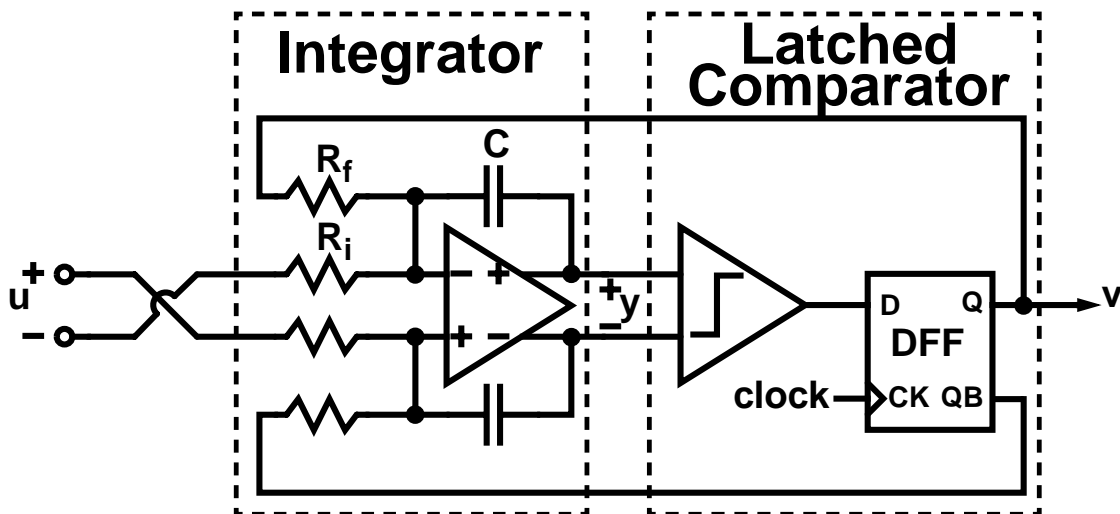


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## CT Implementation of MOD1

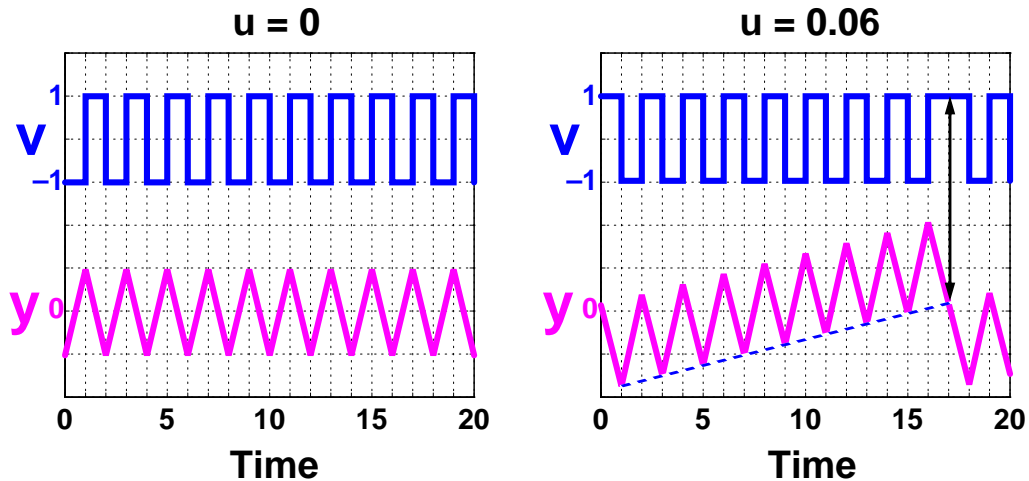
- $R_i/R_f$  sets the full-scale;  $C$  is arbitrary  
 Also observe that an input at  $f_s$  is rejected by the integrator— *inherent anti-aliasing*



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# MOD1-CT Waveforms



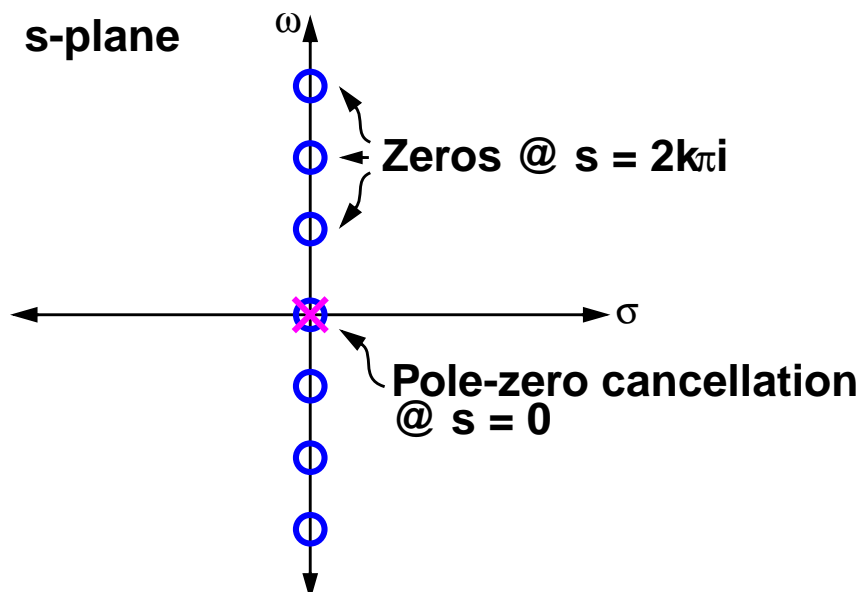
- With  $u=0$ ,  $v$  alternates between +1 and -1
- With  $u>0$ ,  $y$  drifts upwards;  $v$  contains consecutive +1s to counteract this drift

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$$\text{MOD1-CT STF} = \frac{1 - z^{-1}}{s}$$

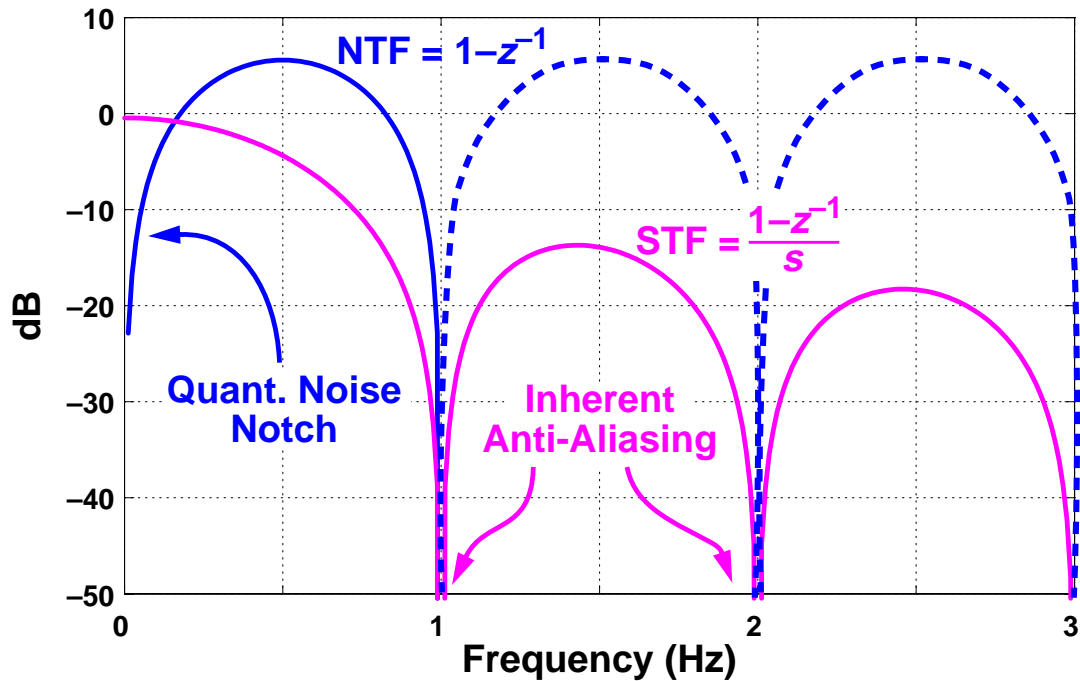
Recall  $z = e^s$



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# MOD1-CT Frequency Responses



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## Summary

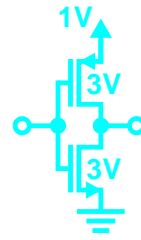
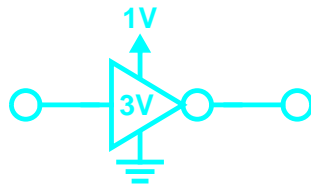
- $\Delta\Sigma$  works by spectrally separating the quantization noise from the signal  
Requires oversampling.  $OSR \equiv f_s / (2f_B)$ .
- Noise-shaping is achieved by the use of *filtering* and *feedback*
- A binary DAC is *inherently linear*, and thus a binary  $\Delta\Sigma$  modulator is too
- MOD1 has  $NTF(z) = 1 - z^{-1}$   
 $\Rightarrow$  Arbitrary accuracy for DC inputs.  
1.5 bit/octave SQNR-OSR trade-off.
- MOD1-CT has *inherent anti-aliasing*

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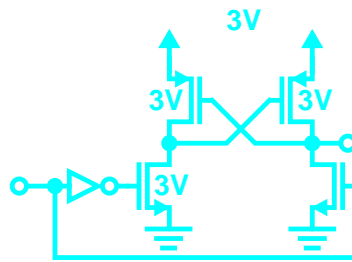
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# NLCOTD

3V → 1V:



1V → 3V:

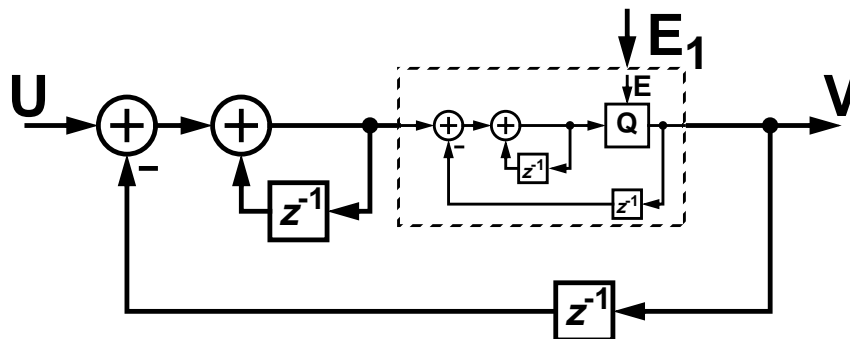


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## 3. MOD2: 2<sup>nd</sup>-Order $\Delta\Sigma$ Modulator [Ch. 3 of Schreier & Temes]

- Replace the quantizer in MOD1 with another copy of MOD1 in a recursive fashion:



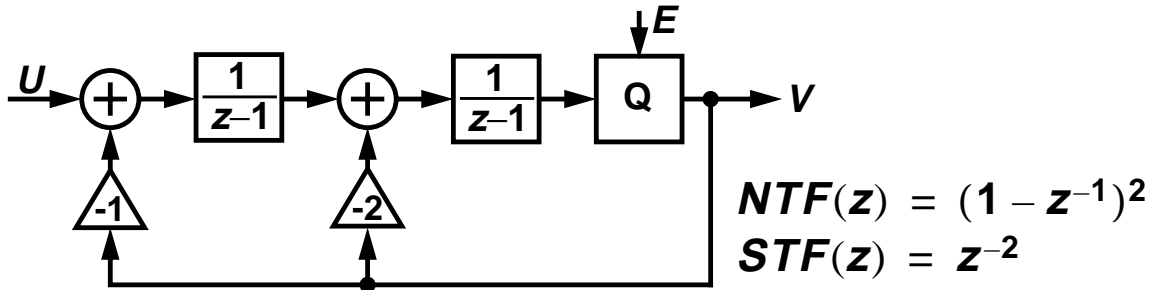
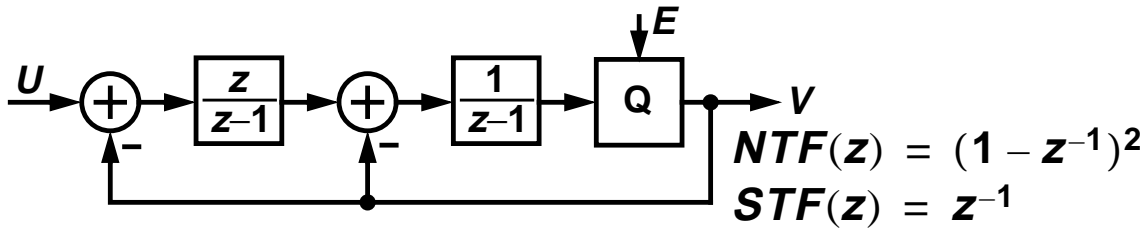
$$V(z) = U(z) + (1-z^{-1})E_1(z), \quad E_1(z) = (1-z^{-1})E(z)$$

$$\Rightarrow V(z) = U(z) + (1-z^{-1})^2 E(z)$$

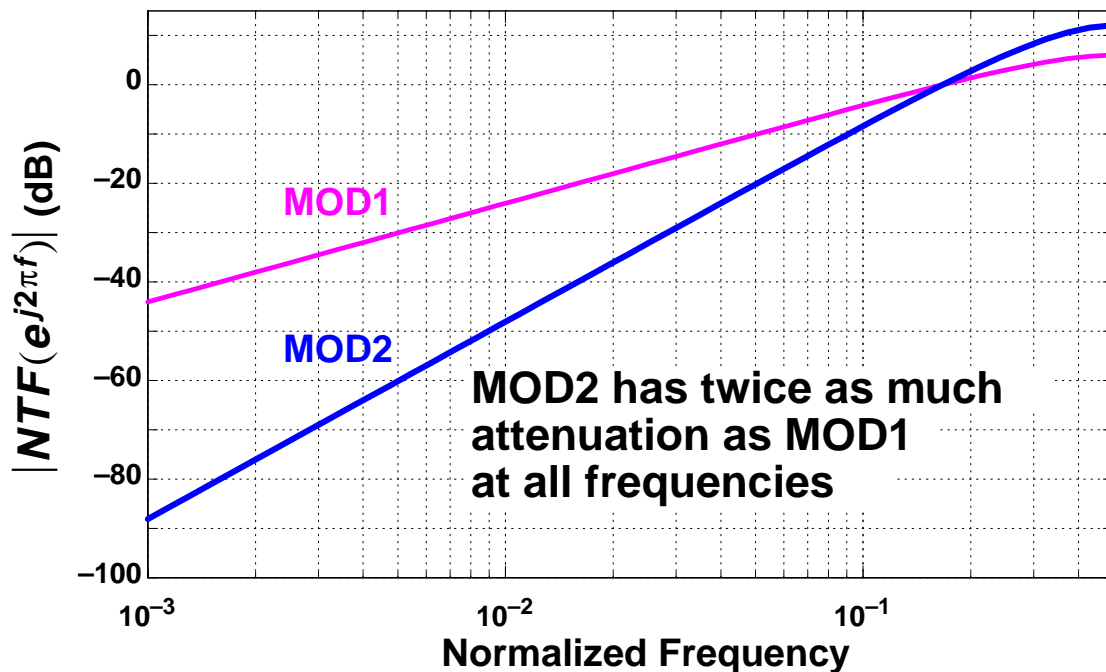
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# Simplified Block Diagrams



# NTF Comparison



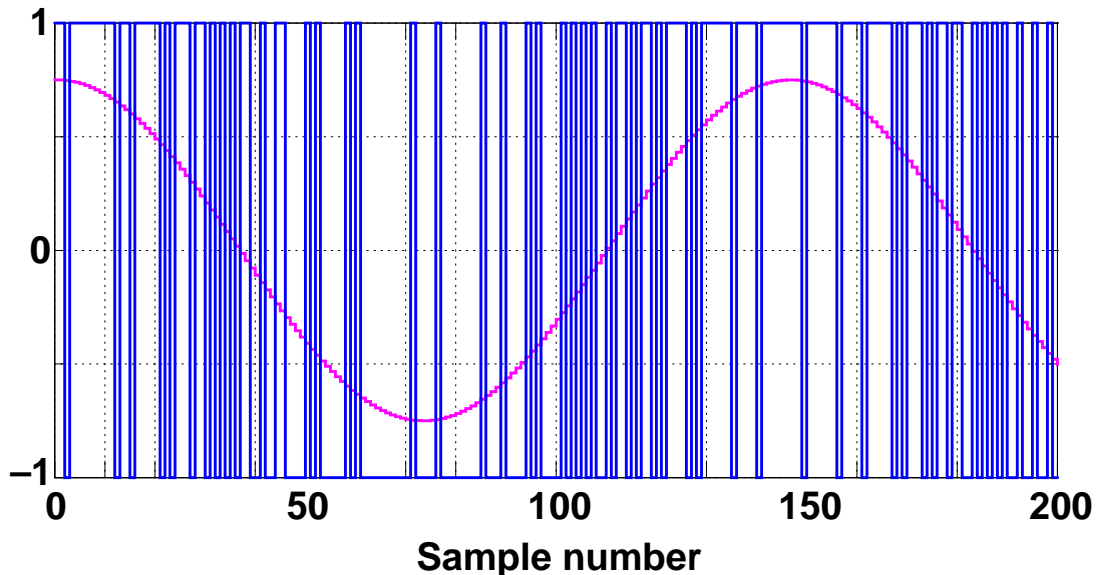
# In-band Quant. Noise Power

- For MOD2,  $|H(e^{j\omega})|^2 \approx \omega^4$
- As before,  $IQNP = \int_0^{\omega_B} |H(e^{j\omega})|^2 S_{ee}(\omega) d\omega$  and  $S_{ee}(\omega) = \sigma_e^2/\pi$
- So now  $IQNP = \frac{\pi^4 \sigma_e^2}{5} (OSR)^{-5} \times$   
With binary quantization to  $\pm 1$ ,  
 $\Delta = 2$  and thus  $\sigma_e^2 = \Delta^2/12 = 1/3$ .
- “An octave increase in *OSR* increases MOD2’s SQNR by 15 dB (2.5 bits)”

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## Simulation Example Input at 75% of FullScale

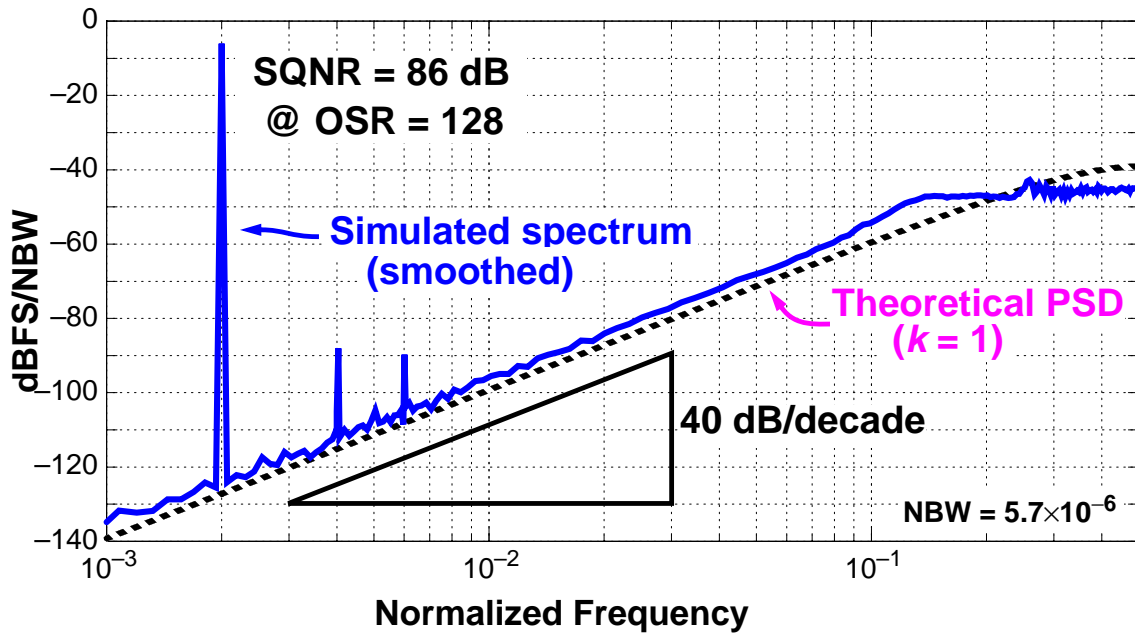


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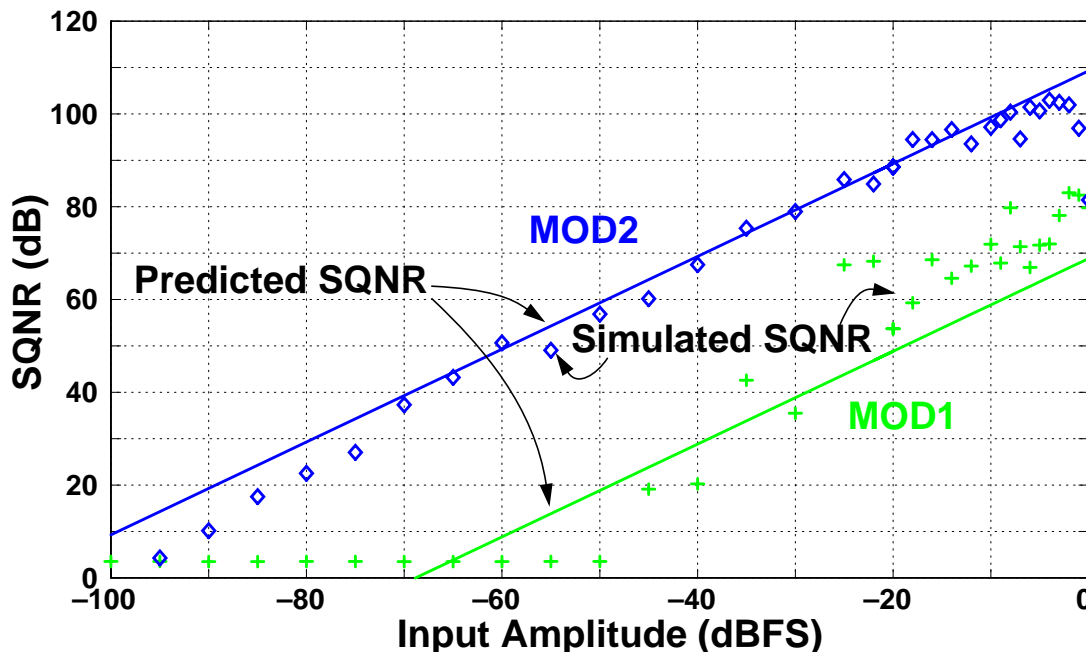
# Simulated MOD2 PSD

## Input at 50% of FullScale



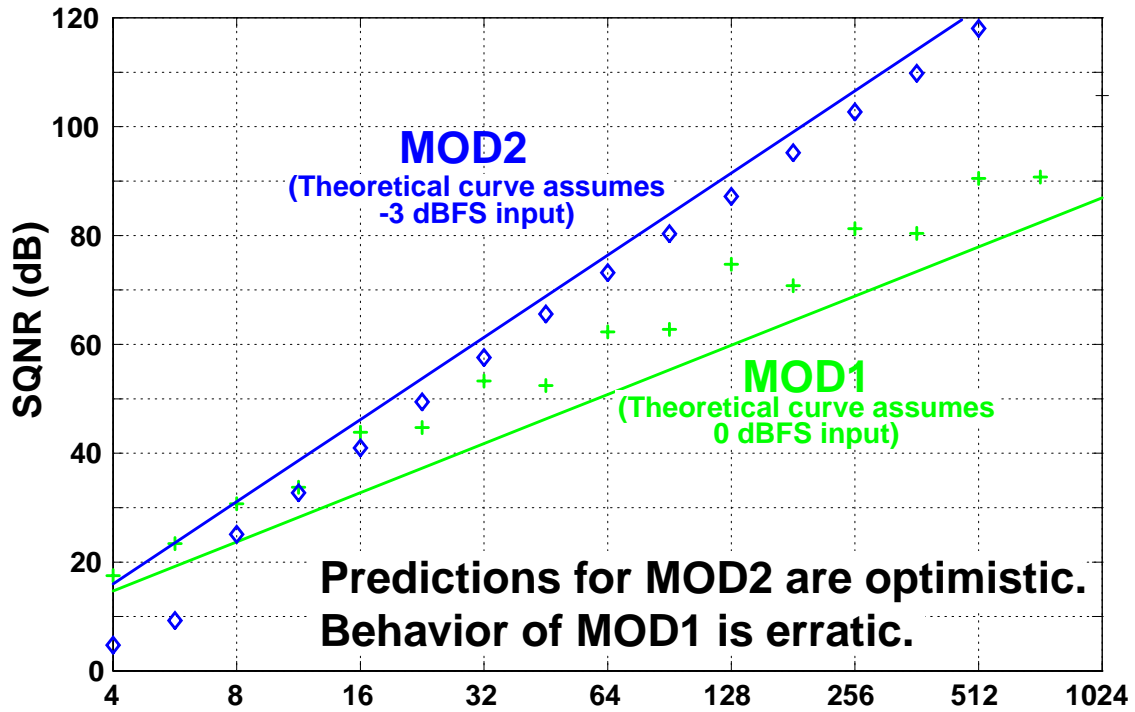
# SQNR vs. Input Amplitude

## MOD1 & MOD2 @ OSR = 256

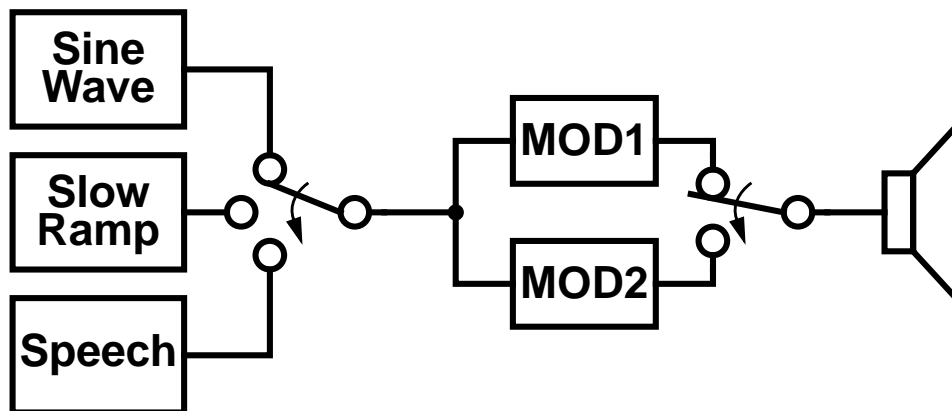




# SQNR vs. OSR



## Audio Demo: MOD1 vs. MOD2 [dsdemo4]



# MOD1 + MOD2 Summary

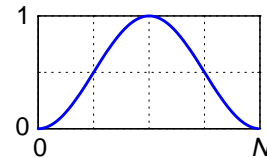
- $\Delta\Sigma$  ADCs rely on filtering and feedback to achieve high SNR despite coarse quantization  
They also rely on digital signal processing.  
 $\Delta\Sigma$  ADCs need to be followed by a digital decimation filter and  $\Delta\Sigma$  DACs need to be preceded by a digital interpolation filter.
- Oversampling eases analog filtering requirements  
Anti-alias filter in an ADC; image filter in a DAC.
- Binary quantization yields inherent linearity
- MOD2 is better than MOD1  
15 dB/octave vs. 9 dB/octave SQNR-OSR trade-off.  
Quantization noise more white.  
Higher-order modulators are even better.

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## 4. Good FFT Practice [Appendix A of Schreier & Temes]

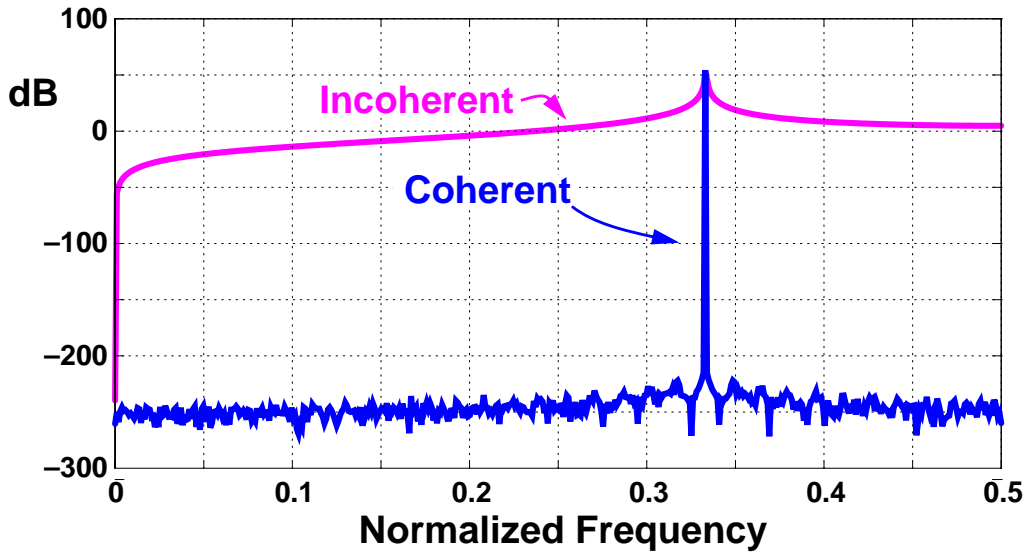
- Use coherent sampling  
I.e. have an integer number of cycles in the record.
- Use windowing  
A Hann window  $w(n) = (1 - \cos(2\pi n/N))/2$  works well.
- Use enough points  
Recommend  $N = 64 \cdot OSR$ .
- Scale (and smooth) the spectrum  
A full-scale sine wave should yield a 0-dBFS peak.
- State the noise bandwidth  
For a Hann window,  $NBW = 1.5/N$ .



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# Coherent vs. Incoherent Sampling



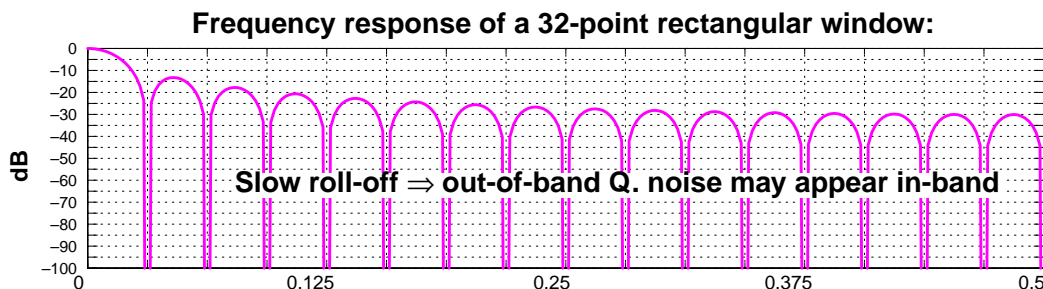
- Coherent sampling: only one non-zero FFT bin
- Incoherent sampling: “spectral leakage”

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## 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”

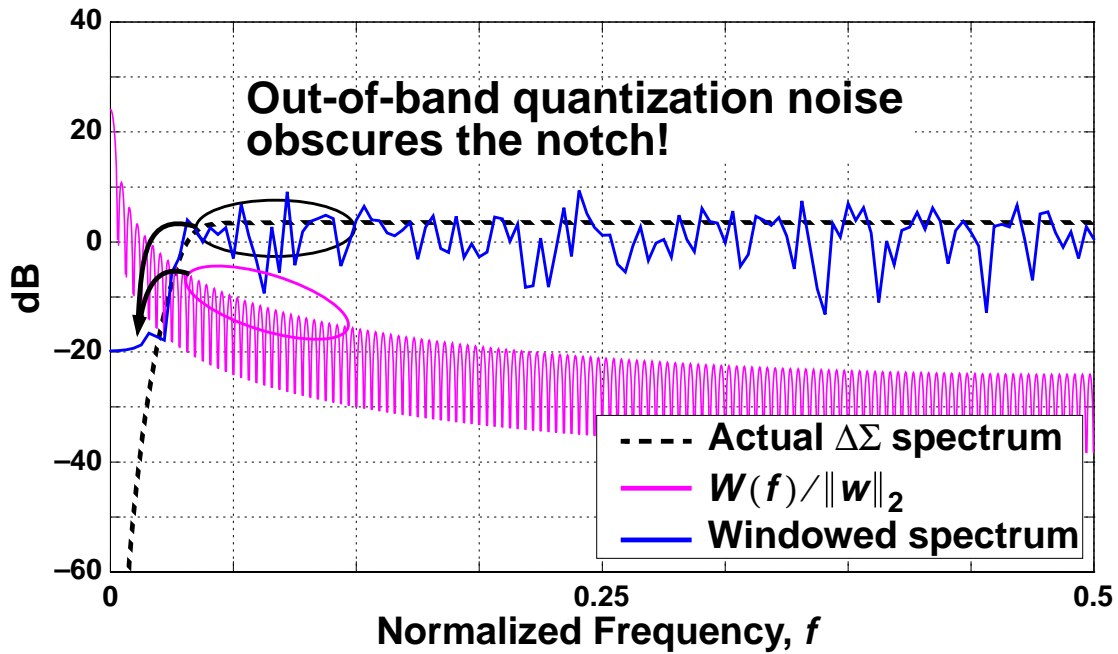


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# Example Spectral Disaster

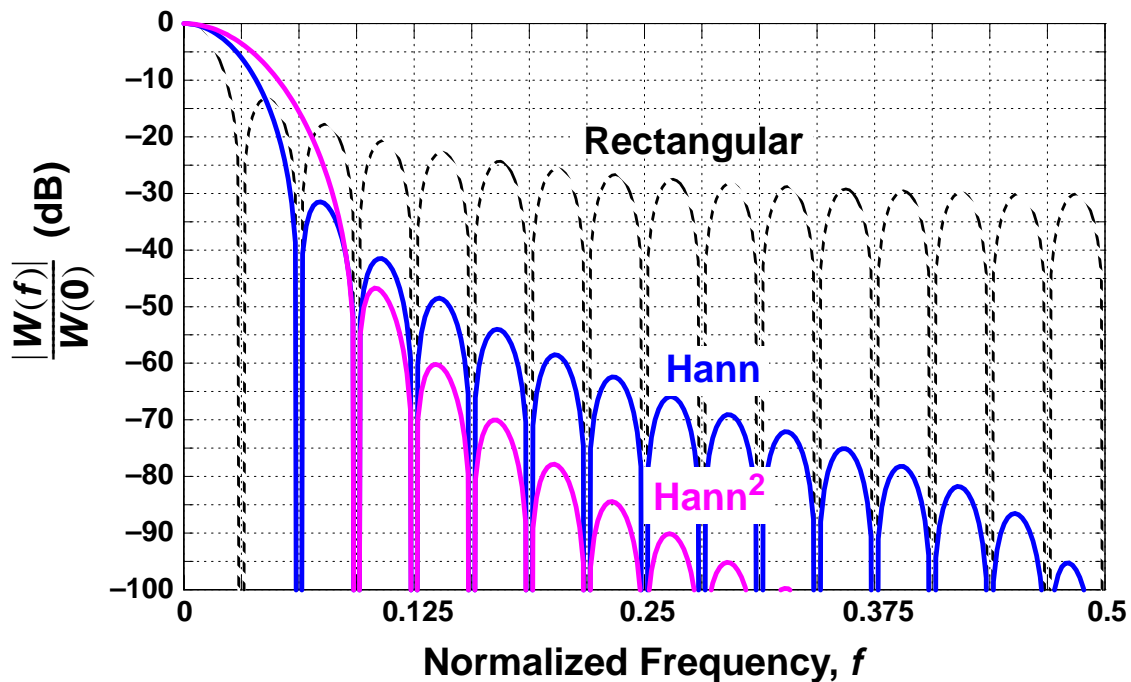
## Rectangular window, $N = 256$



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## Window Comparison ( $N = 16$ )



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# 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
$\ w\ _2^2 = \sum w(n)^2$	$N$	$3N/8$	$35N/128$
$W(0) = \sum w(n)$	$N$	$N/2$	$3N/8$
$NBW = \frac{\ w\ _2^2}{W(0)^2}$	$1/N$	$1.5/N$	$35/18N$

†. MATLAB's "hann" function causes spectral leakage of tones located in FFT bins unless you add the optional argument "periodic."

## 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

# FFT Scaling

- The FFT implemented in MATLAB is

$$X_M(k+1) = \sum_{n=0}^{N-1} x_M(n+1) e^{-j\frac{2\pi kn}{N}}$$

- If  $x(n) = A \sin(2\pi fn/N)^\dagger$ , then

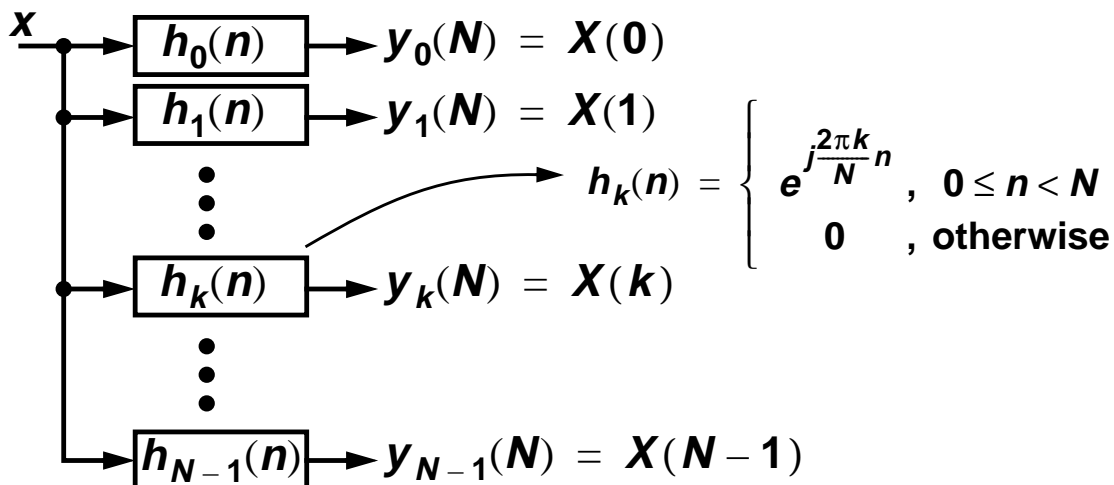
$$|X(k)| = \begin{cases} \frac{AN}{2} & , k = f \text{ or } N - f \\ 0 & , \text{otherwise} \end{cases}$$

⇒ Need to divide FFT by  $(N/2)$  to get  $A$ .

†.  $f$  is an integer in  $(0, N/2)$ . I've defined  $X(k) \equiv X_M(k+1)$ ,  $x(n) \equiv x_M(n+1)$  since Matlab indexes from 1 rather than 0.

# The Need For Smoothing

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



⇒ an FFT yields a high-variance spectral estimate

# How To Do Smoothing

## 1 Average multiple FFTs

Implemented by MATLAB's `psd()` function

## 2 Take one big FFT and “filter” the spectrum

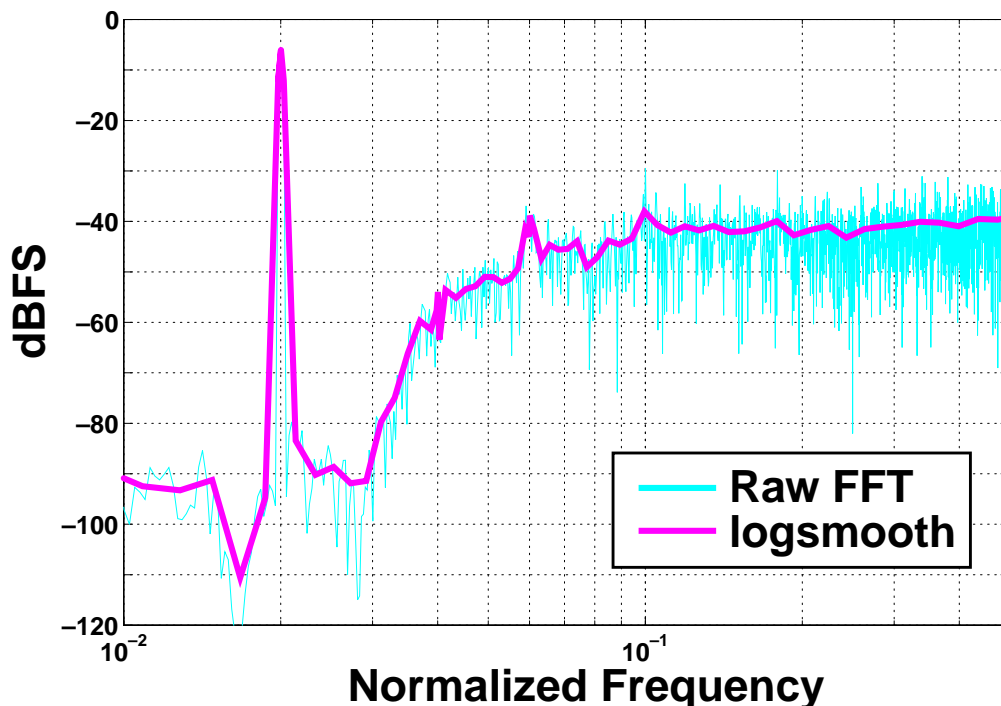
Implemented by the  $\Delta\Sigma$  Toolbox's `logsmooth()` function

`logsmooth()` averages an exponentially-increasing number of bins in order to reduce the density of points in the high-frequency regime and make a nice log-frequency plot

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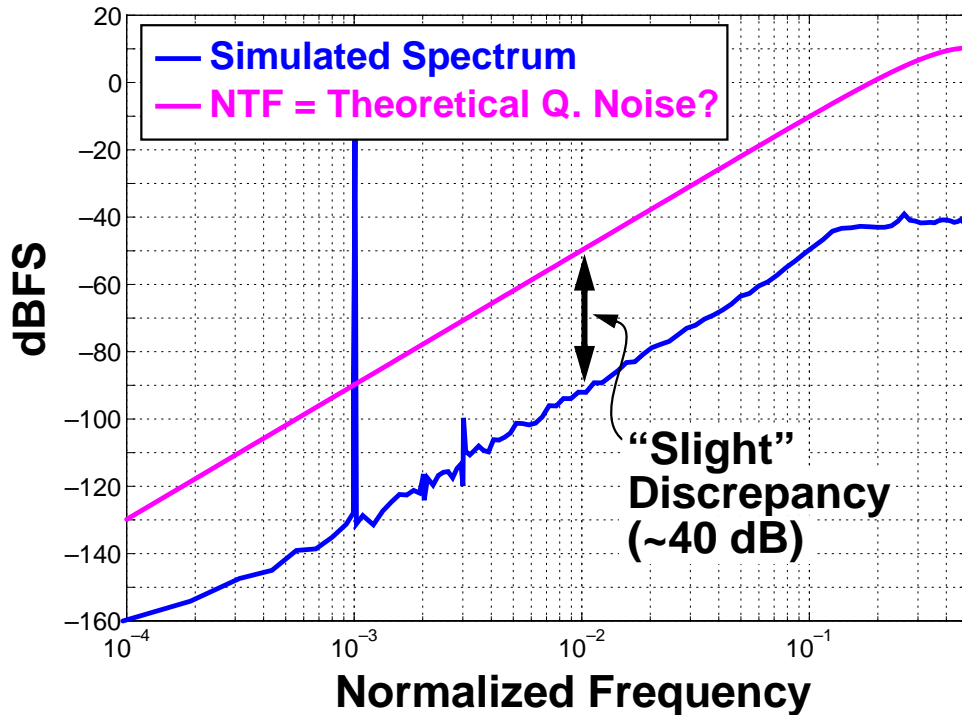
## Raw and Smoothed Spectra



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# Simulation vs. Theory (MOD2)



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## What Went Wrong?

- 1 We normalized the spectrum so that a full-scale sine wave (which has a power of 0.5) comes out at 0 dB (whence the “dBFS” units)

⇒ We need to do the same for the error signal.  
i.e. use  $S_{ee}(f) = 4/3$ .

But this makes the discrepancy 3 dB worse.

- 2 We tried to plot a *power spectral density* together with something that we want to interpret as a *power spectrum*
- Sine-wave components are located in individual FFT bins, but broadband signals like noise have their power spread over all FFT bins!

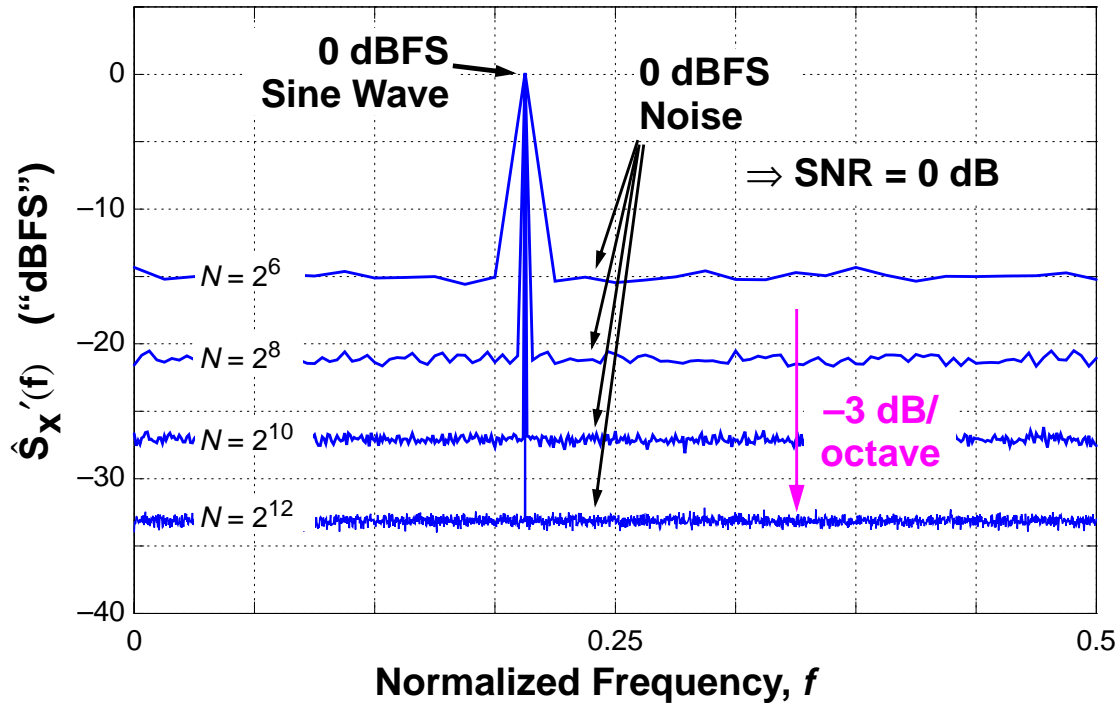
The “noise floor” depends on the length of the FFT.

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# Spectrum of a Sine Wave + Noise



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## Observations

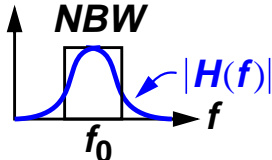
- The power of the sine wave is given by the height of its spectral peak
- The power of the noise is spread over all bins  
The greater the number of bins, the less power there is in any one bin.
- Doubling  $N$  reduces the power per bin by a factor of 2 (i.e. 3 dB)  
But the total integrated noise power does *not* change.

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# So How Do We Handle Noise?

- Recall that an FFT is like a filter bank
- The longer the FFT, the narrower the bandwidth of each filter and thus the lower the power at each output
- We need to know the *noise bandwidth* (NBW) of the filters in order to convert the power in each bin (filter output) to a power density
- For a filter with frequency response  $H(f)$ ,

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


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## FFT Noise Bandwidth Rectangular Window

$$h_k(n) = \exp\left(j\frac{2\pi k}{N}n\right), H_k(f) = \sum_{n=0}^{N-1} h_k(n) \exp(-j2\pi fn)$$

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

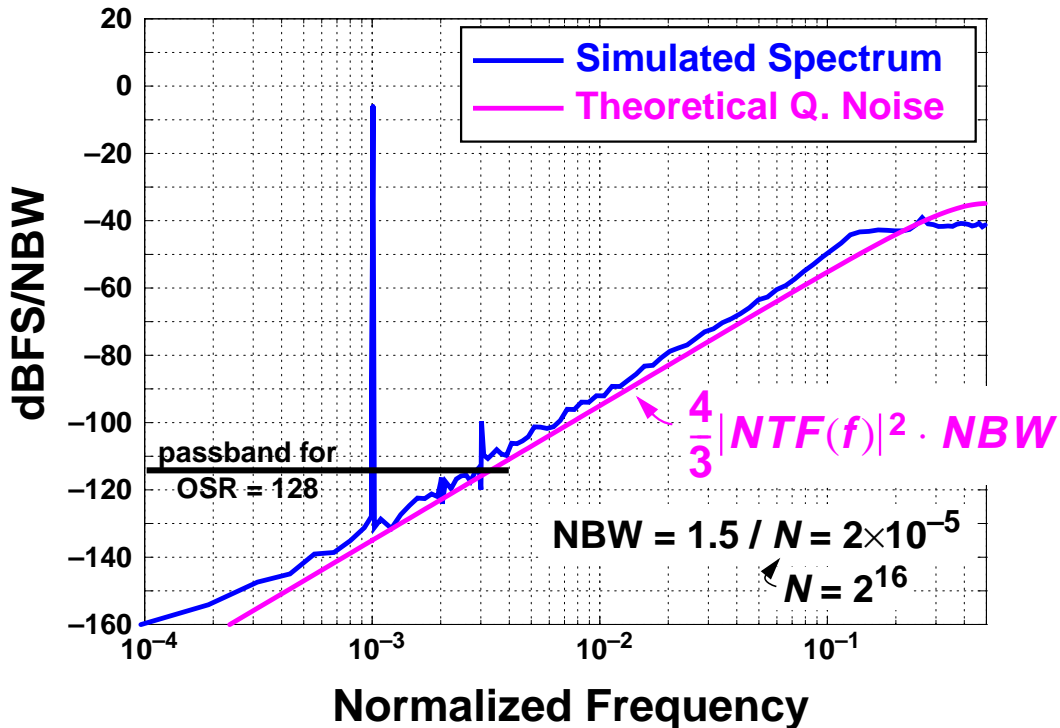
$$\int |H_k(f)|^2 = \sum |h_k(n)|^2 = N \text{ [Parseval]}$$

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

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# Better Spectral Plot



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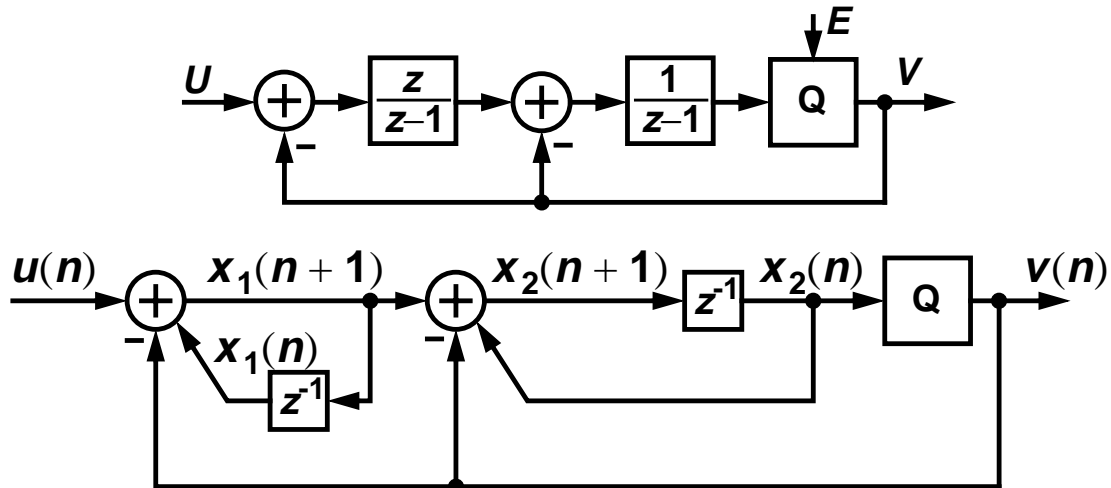
## Homework #1 (Due 2015-01-12)

- A. Create a Matlab function that computes MOD1's output sequence given a vector of input samples and exercise your function in the following ways.
- 1 Verify that the average of the output equals the input for a few random DC inputs in  $[-1,1]$ .
  - 2 Plot the output spectrum with a half-scale sine-wave input. Use good FFT practice. Include the theoretical quantization noise curve and list the theoretical and simulated SQNR for OSR = 128.
- B. Repeat with MOD2.
- C. Compose your own question and answer it.

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# MOD2 Expanded



Difference Equations:

$$v(n) = Q(x_2(n))$$

$$x_1(n+1) = x_1(n) - v(n) + u(n)$$

$$x_2(n+1) = x_2(n) - v(n) + x_1(n+1)$$

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## Example Matlab Code

```
function [v] = simulateMOD2(u)
    x1 = 0;
    x2 = 0;
    for i = 1:length(u)
        v(i) = quantize( x2 );
        x1 = x1 + u(i) - v(i);
        x2 = x2 + x1 - v(i);
    end
    return
```

```
function v = quantize( y )
    if y >= 0
        v = 1;
    else
        v = -1;
    end
    return
```

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# **What You Learned Today**

## **And what the homework should solidify**

- 1 MOD1: 1<sup>st</sup>-order  $\Delta\Sigma$  modulator**  
Structure and theory of operation
- 2 Inherent linearity of binary modulators**
- 3 Inherent anti-aliasing of continuous-time modulators**
- 4 MOD2: 2<sup>nd</sup>-order  $\Delta\Sigma$  modulator**
- 5 Good FFT practice**