# ESE 531: Digital Signal Processing

Lec 12: February 25, 2020 Data Converters, Noise Shaping



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- Data Converters
  - Anti-aliasing
  - ADC
    - Quantization
  - Practical DAC
- Noise Shaping

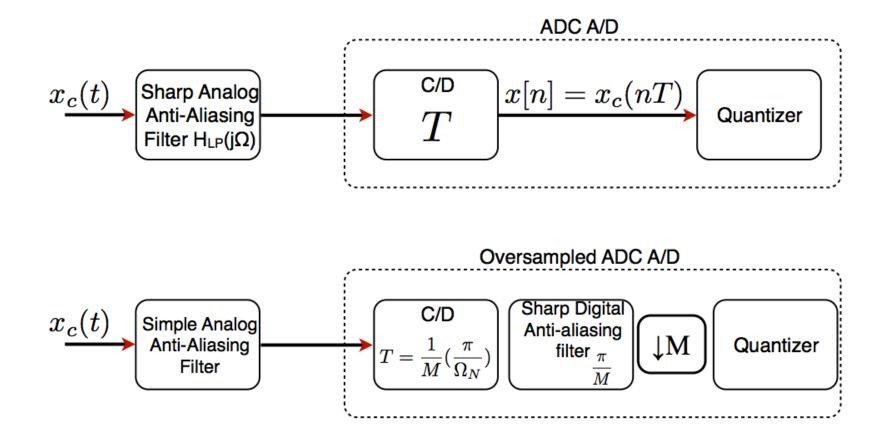
## ADC

## Analog to Digital Converter

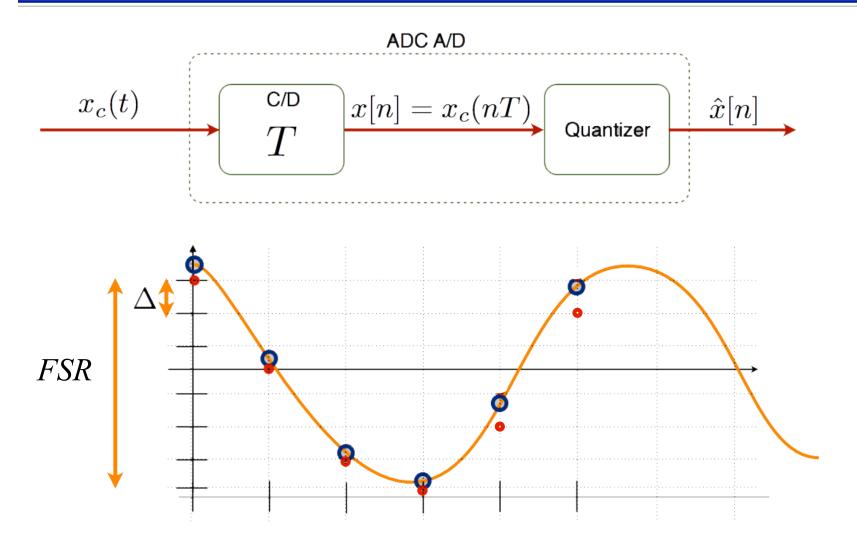


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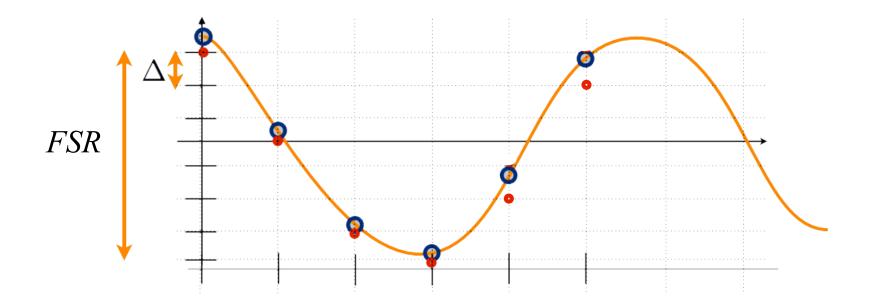




Sampling and Quantization

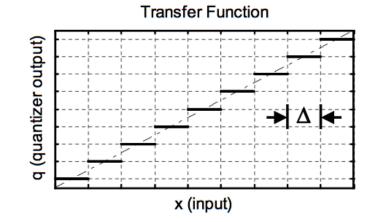
• For an input signal with  $V_{pp}$ =FSR with B bits

 $\Delta = \frac{FSR}{2^B}$ 



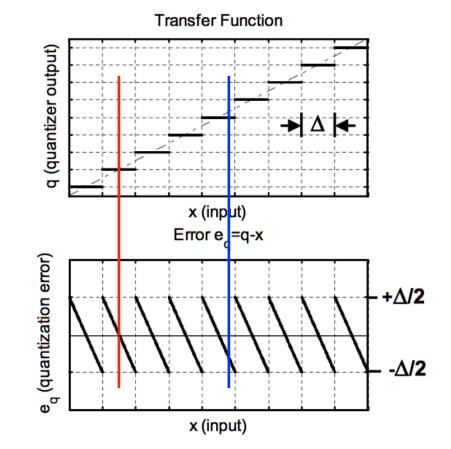


#### $\Box$ Quantization step $\Delta$



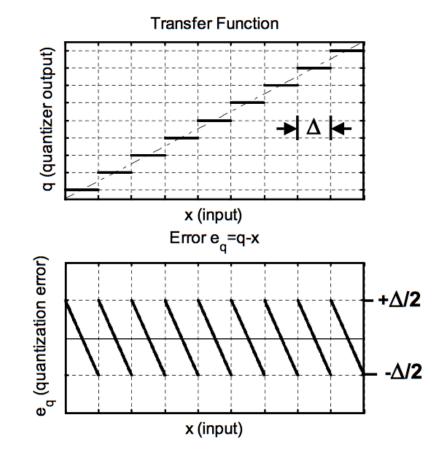


- $\Box$  Quantization step  $\Delta$
- Quantization error has sawtooth shape
  - Bounded by  $-\Delta/2$ ,  $+\Delta/2$



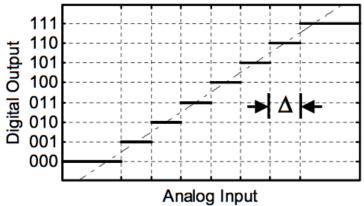


- **Quantization step**  $\Delta$
- Quantization error has sawtooth shape
  - Bounded by  $-\Delta/2$ ,  $+\Delta/2$
- Ideally infinite input range and infinite number of quantization levels





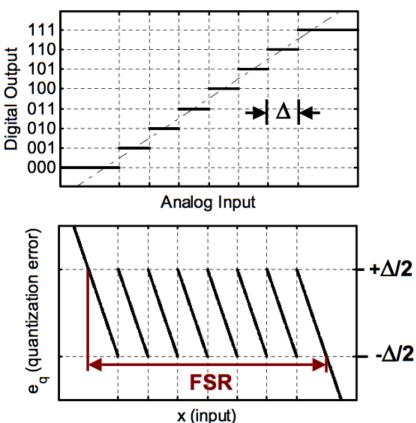
- Practical quantizers have a limited input range and a finite set of output codes
- E.g. a 3-bit quantizer can map onto
   2<sup>3</sup>=8 distinct output codes



Ideal B-bit Quantizer

- Practical quantizers have a limited input range and a finite set of output codes
- E.g. a 3-bit quantizer can map onto
   2<sup>3</sup>=8 distinct output codes

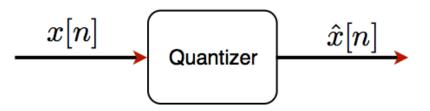
- Quantization error grows out of bounds beyond code boundaries
- We define the full scale range (FSR) as the maximum input range that satisfies  $|e_q| \le \Delta/2$ 
  - Implies that  $FSR = 2^B \cdot \Delta$



# Effect of Quantization Error on Signal

- Quantization error is a deterministic function of the signal
  - Consequently, the effect of quantization strongly depends on the signal itself
- Unless, we consider fairly trivial signals, a deterministic analysis is usually impractical
  - More common to look at errors from a statistical perspective
  - "Quantization noise"
- □ Two aspects
  - How much noise power (variance) does quantization add to our samples?
  - How is this noise distributed in frequency?

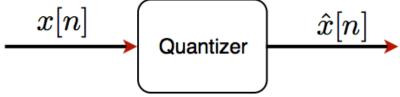




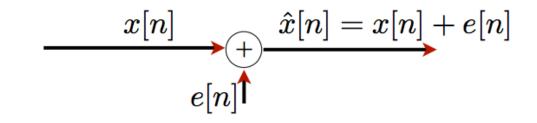
Model quantization error as noise:

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Model quantization error as noise:

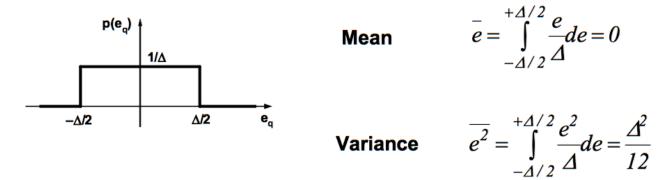


□ In that case:

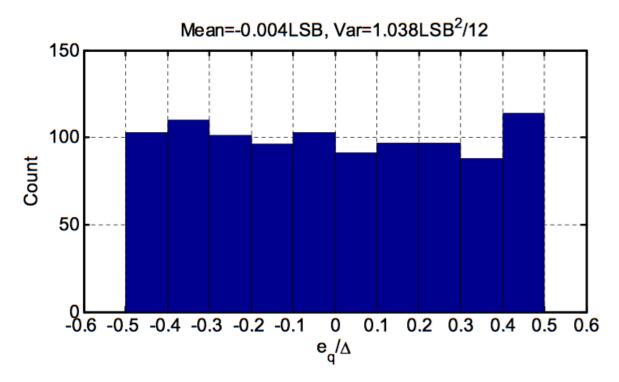
 $-\Delta/2 \leq e[n] < \Delta/2$ 

Quantization Error Statistics

- Crude assumption: e<sub>q</sub>(x) has uniform probability density
- This approximation holds reasonably well in practice when
  - Signal spans large number of quantization steps
  - Signal is "sufficiently active"
  - Quantizer does not overload



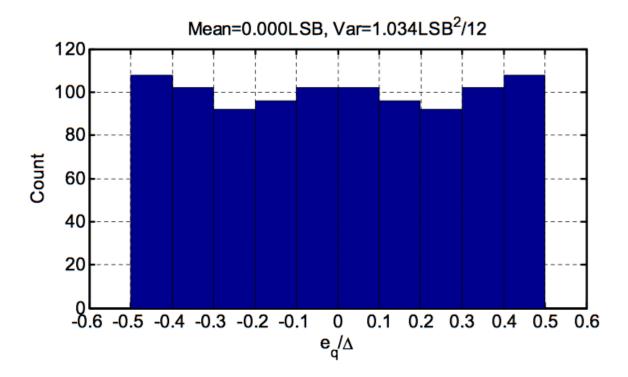




• Shown is a histogram of  $e_q$  in an 8-bit quantizer

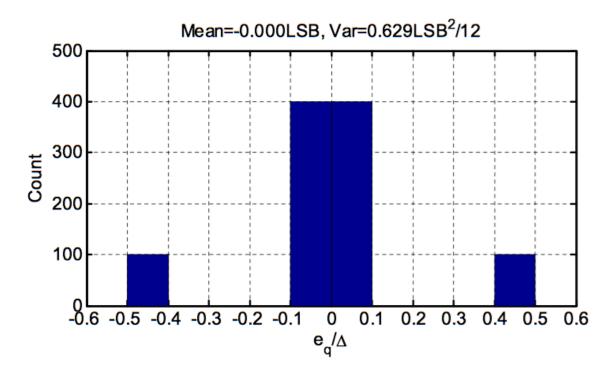
• Input sequence consists of 1000 samples with Gaussian distribution,  $4\sigma = FSR$ 





□ Same as before, but now using a sinusoidal input signal with  $f_{sig}/f_s = 101/1000$ 





- □ Same as before, but now using a sinusoidal input signal with  $f_{sig}/f_s = 100/1000$
- □ What went wrong?



$$\mathbf{v}_{\rm sig}[n] = \cos\left(2\pi \cdot \frac{f_{\rm sig}}{f_{\rm S}} \cdot n\right)$$

□ Signal repeats every m samples, where m is the smallest integer that satisfies  $m \cdot \frac{f_{sig}}{f_S} = \text{integer}$ 



$$\mathbf{v}_{\rm sig}(n) = \cos\left(2\pi \cdot \frac{f_{\rm sig}}{f_{\rm S}} \cdot n\right)$$

• Signal repeats every m samples, where m is the smallest integer that satisfies  $m \cdot \frac{f_{sig}}{f_s} = \text{integer}$  $m \cdot \frac{101}{1000} = \text{integer} \Rightarrow m=1000$  $m \cdot \frac{100}{1000} = \text{integer} \Rightarrow m=10$ 

□ This means that in the last case  $e_q(n)$  consists at best of 10 different values, even though we took 1000 samples

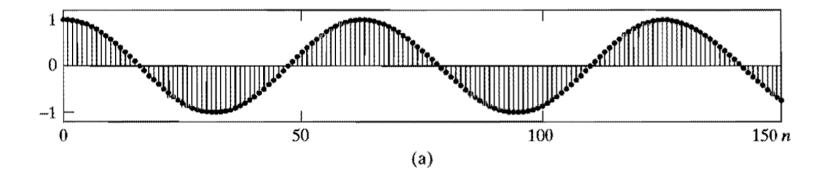
Noise Model for Quantization Error

### • Assumptions:

- Model e[n] as a sample sequence of a stationary random process
- e[n] is not correlated with x[n]
- e[n] not correlated with e[m] where  $m \neq n$
- $e[n] \sim U[-\Delta/2, \Delta/2]$  (uniform pdf)
- Result:
  Variance is: \sigma\_e^2 = \frac{\Delta^2}{12}
- Assumptions work well for signals that change rapidly, are not clipped, and for small  $\Delta$

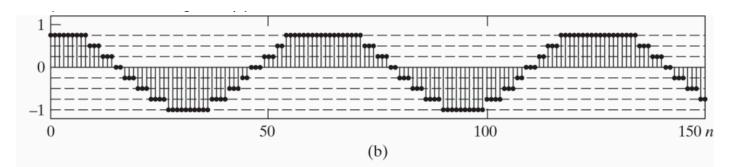


Figure 4.57 Example of quantization noise. (a) Unquantized samples of the signal x[n] = 0.99cos(n/10).





• Figure 4.57(continued) (b) Quantized samples of the cosine waveform in part (a) with a 3-bit quantizer.





• **Figure 4.57**(continued) (b) Quantized samples of the cosine waveform in part (a) with a 3bit quantizer. (c) Quantization error sequence for 3-bit quantization of the signal in (a).

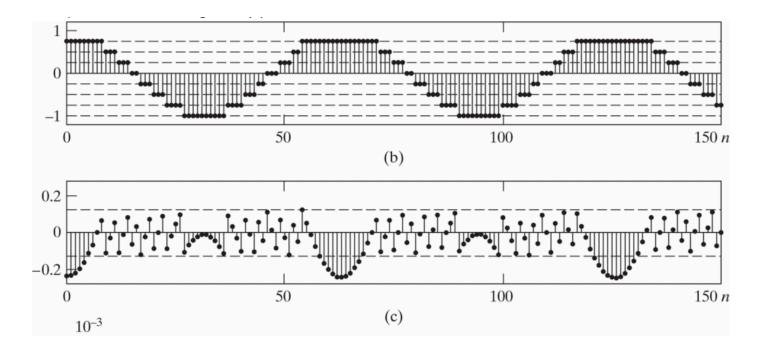
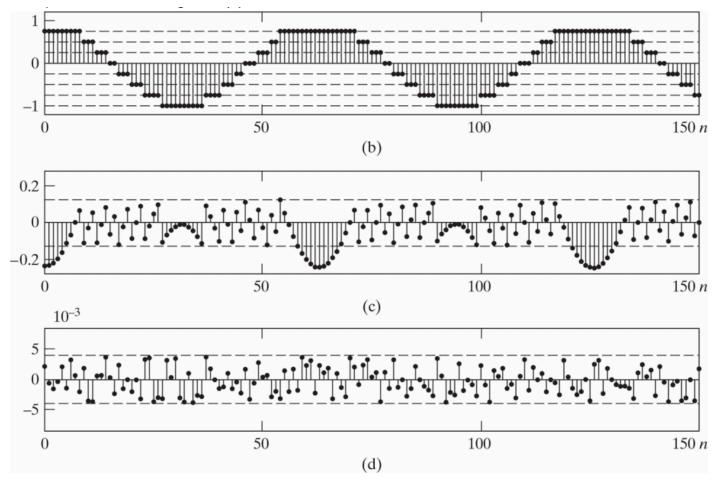




Figure 4.57(continued) (b) Quantized samples of the cosine waveform in part (a) with a 3-bit quantizer. (c) Quantization error sequence for 3-bit quantization of the signal in (a). (d) Quantization error sequence for 8-bit quantization of the signal in (a).



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### • For uniform B bits quantizer

$$SNR_Q = 10 \log_{10} \left( \frac{\sigma_x^2}{\sigma_e^2} \right)$$

Signal-to-Quantization-Noise Ratio

### • For uniform B bits quantizer

$$SNR_Q = 10 \log_{10} \left(\frac{\sigma_x^2}{\sigma_e^2}\right)$$
$$= 10 \log_{10} \left(\frac{12 \cdot 2^{2B} \sigma_x^2}{FSR^2}\right)$$

$$\mathrm{SNR}_Q = 6.02B + 10.8 - 20 \log_{10} \left(\frac{FSR}{\sigma_x}\right)^{\text{Quantizer range}}$$



$$SNR_Q = 6.02B + 10.8 - 20 \log_{10} \left(\frac{FSR}{\sigma_x}\right)^{\text{Quantizer range}}$$
rms of amp

- □ Improvement of 6dB with every bit
- The range of the quantization must be adapted to the rms amplitude of the signal
  - Tradeoff between clipping and noise!
  - Often use pre-amp
  - Sometimes use analog auto gain controller (AGC)



• Assuming full-scale sinusoidal input, we have

$$SQNR = \frac{P_{sig}}{P_{qnoise}} =$$

Signal-to-Quantization-Noise Ratio

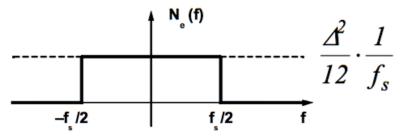
• Assuming full-scale sinusoidal input, we have

$$SQNR = \frac{P_{sig}}{P_{qnoise}} = 6.02B + 1.76 \text{ dB}$$

B (Number of Bits)	SQNR
8	50dB
12	74dB
16	98dB
20	122dB

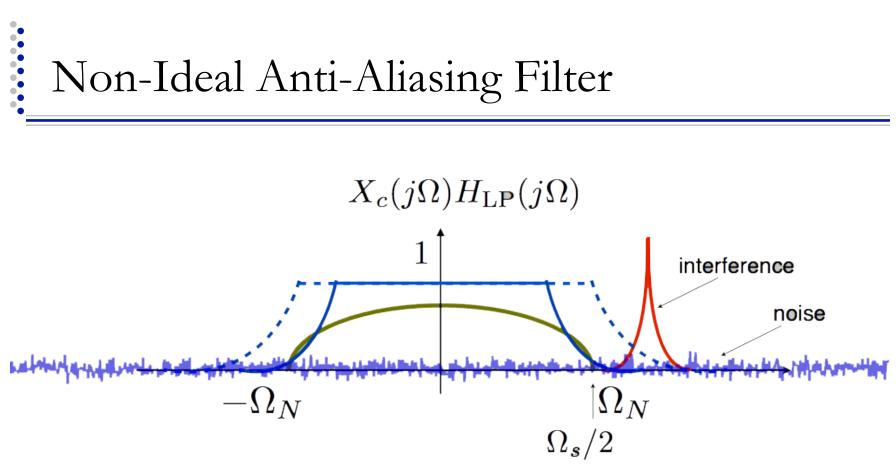
Quantization Noise Spectrum

 If the quantization error is "sufficiently random", it also follows that the noise power is uniformly distributed in frequency



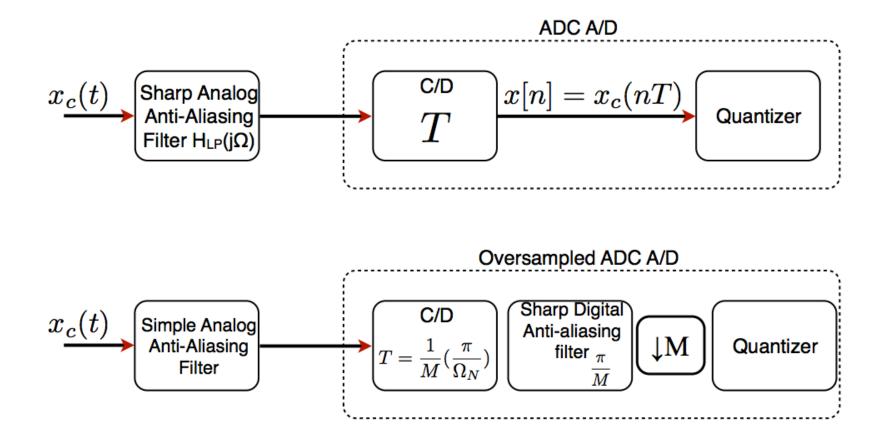
References

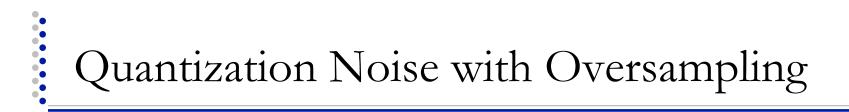
- W. R. Bennett, "Spectra of quantized signals," Bell Syst. Tech. J., pp. 446-72, July 1988.
- B. Widrow, "A study of rough amplitude quantization by means of Nyquist sampling theory," IRE Trans. Circuit Theory, vol. CT-3, pp. 266-76, 1956.

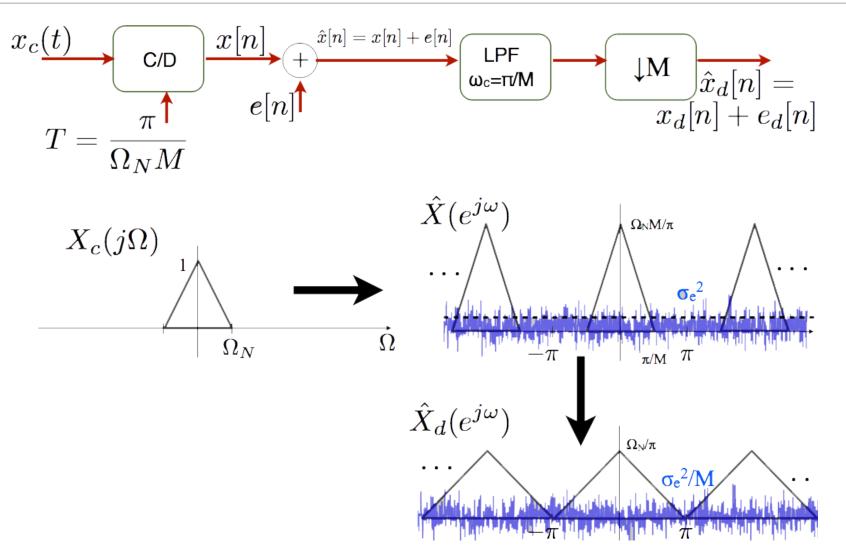


- Problem: Hard to implement sharp analog filter
- Consequence: Crop part of the signal and suffer from noise and interference









Quantization Noise with Oversampling

- Energy of  $x_d[n]$  equals energy of x[n]
  - No filtering of signal!
- □ Noise variance is reduced by factor of M

$$SNR_Q = 6.02B + 10.8 - 20 \log_{10} \left(\frac{FSR}{\sigma_x}\right) + 10 \log_{10} M$$

- For doubling of M we get 3dB improvement, which is the same as 1/2 a bit of accuracy
  - With oversampling of 16 with 8bit ADC we get the same quantization noise as 10bit ADC!

## Practical DAC



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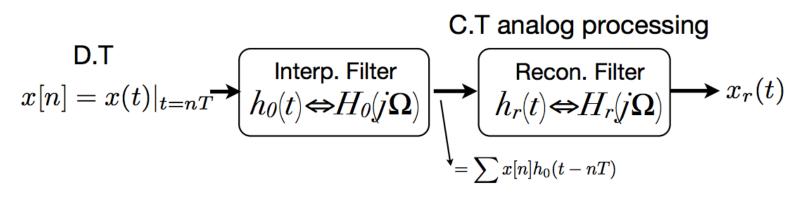
D.T  

$$x[n] = x(t)|_{t=nT}$$
  $\xrightarrow{\text{sinc pulse}}$   $x_r(t) = \sum_{n=-\infty}^{\infty} x[n] \text{sinc} \left(\frac{t-nT}{T}\right)$ 

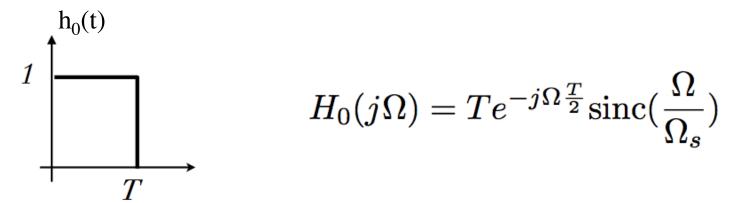
Scaled train of sinc pulses

**Difficult** to generate sinc  $\rightarrow$  Too long!



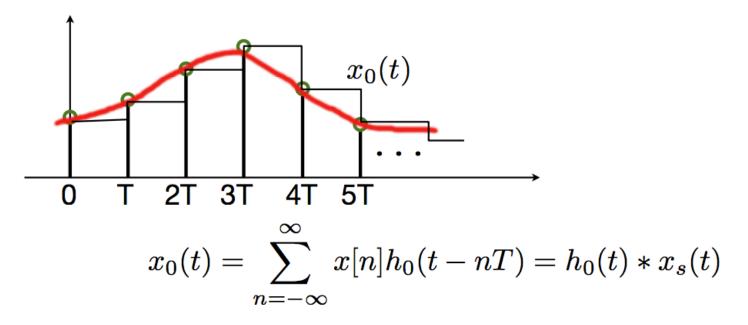


- $h_0(t)$  is finite length pulse  $\rightarrow$  easy to implement
- □ For example: zero-order hold



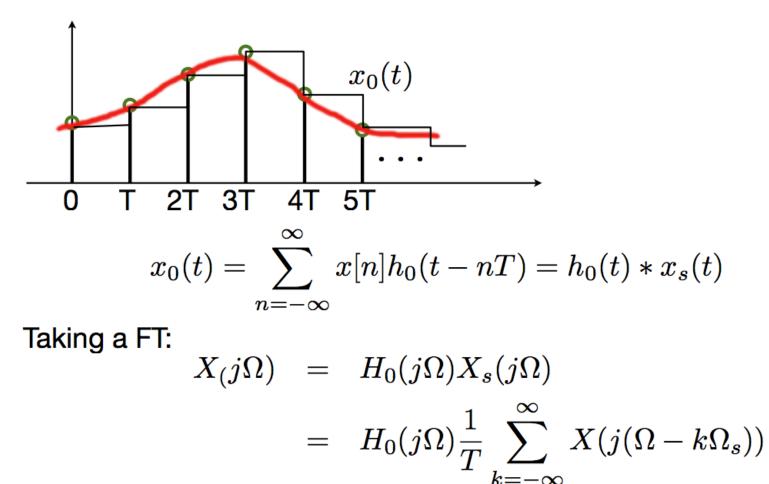


## Zero-Order-Hold interpolation

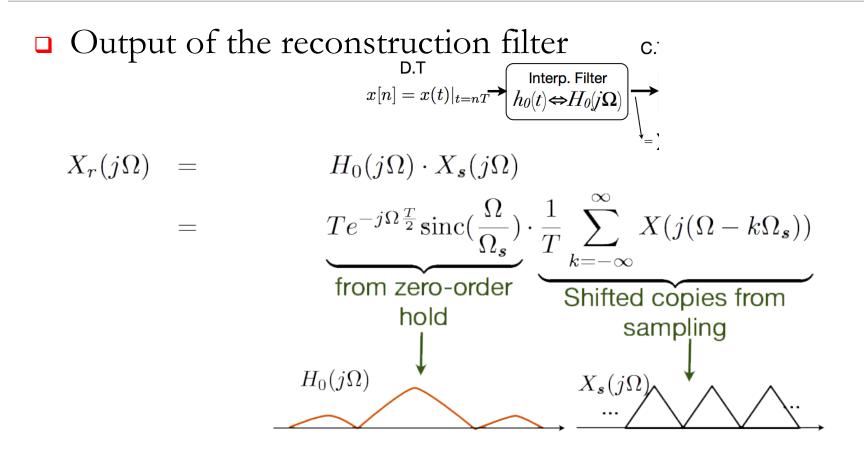




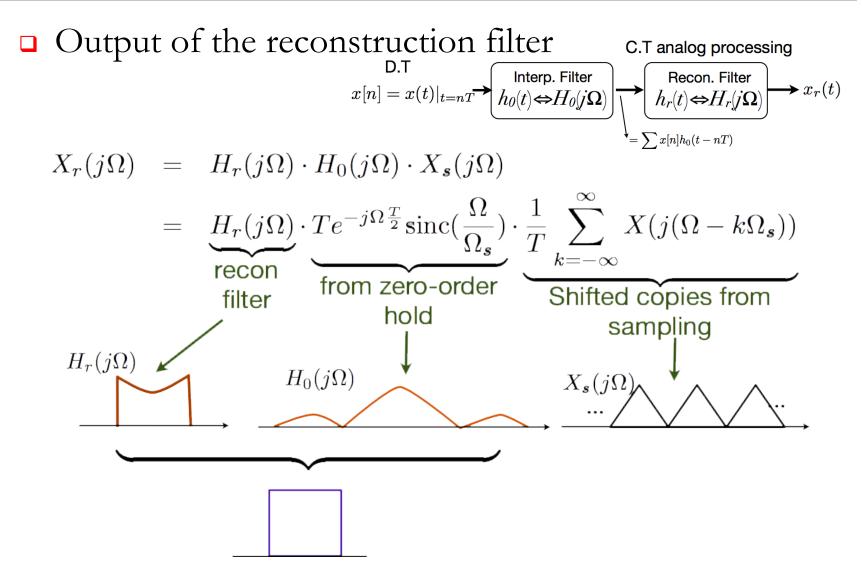
## Zero-Order-Hold interpolation



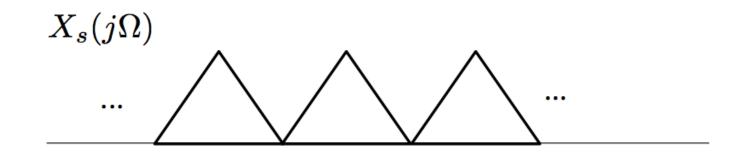
Practical DAC



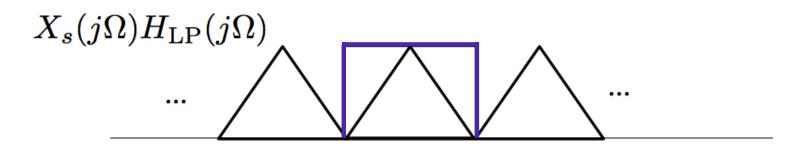




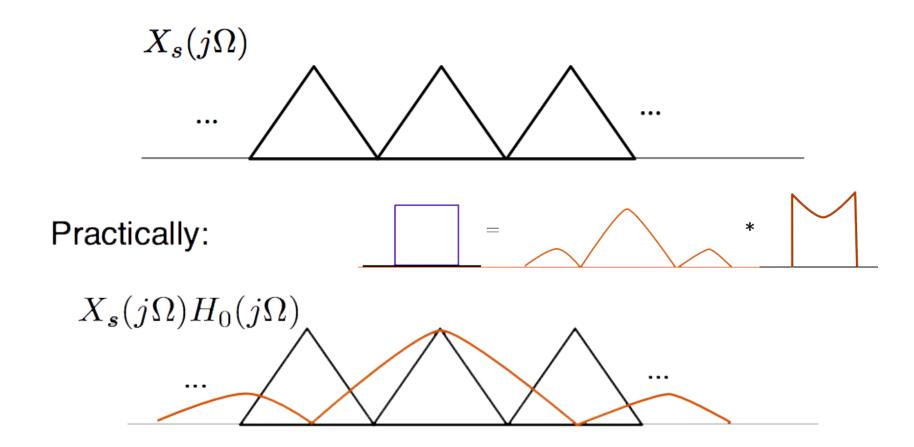




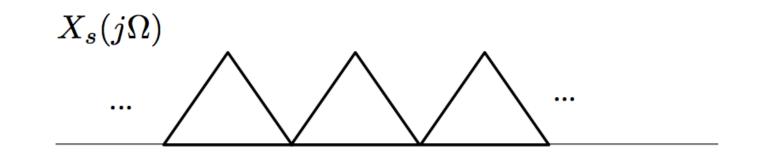
Ideally:



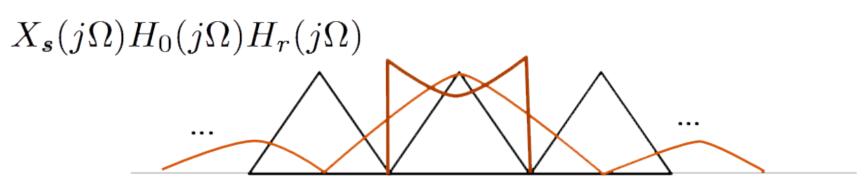




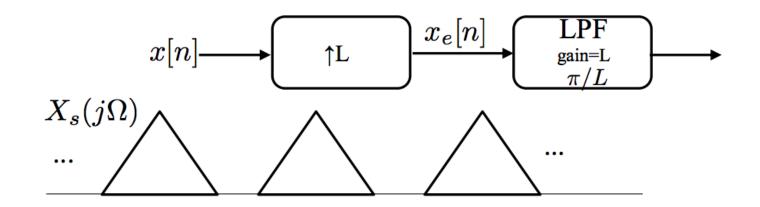




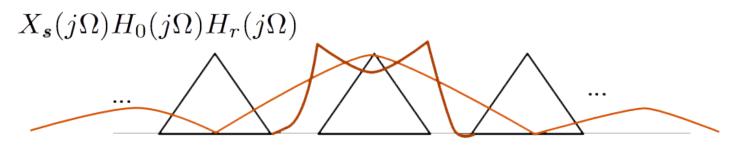
Practically:



Practical DAC with Upsampling



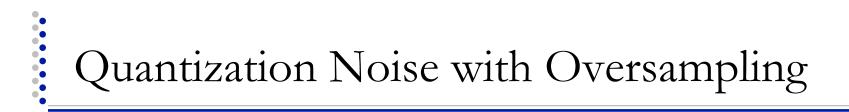
Practically:

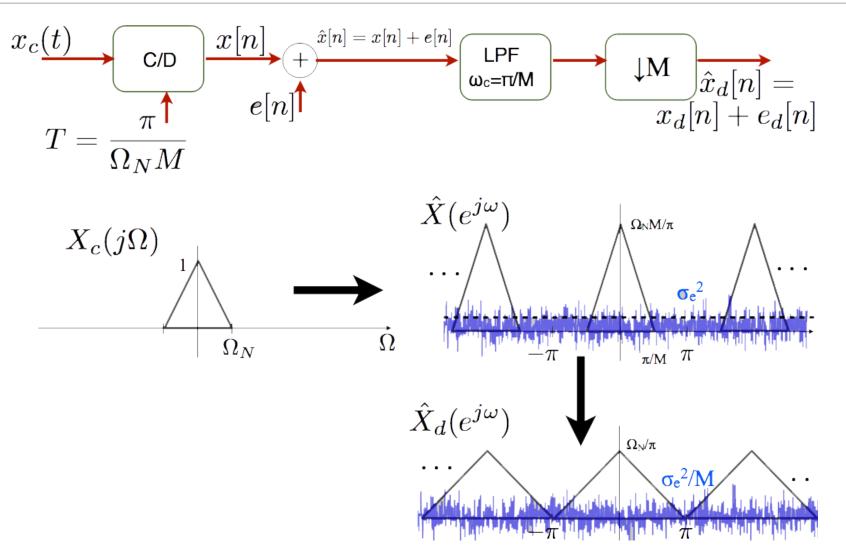


## Noise Shaping



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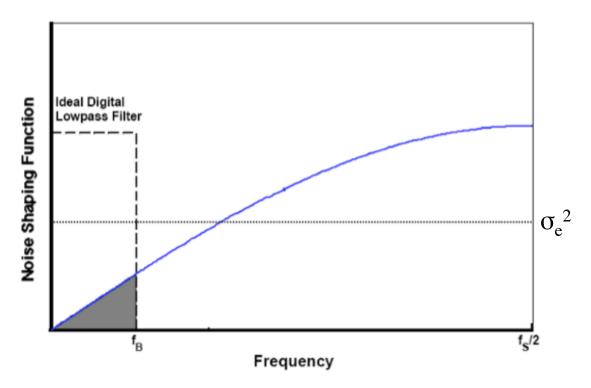
Quantization Noise with Oversampling

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- □ Noise variance is reduced by factor of M

$$SNR_Q = 6.02B + 10.8 - 20\log_{10}\left(\frac{X_m}{\sigma_x}\right) + 10\log_{10}M$$

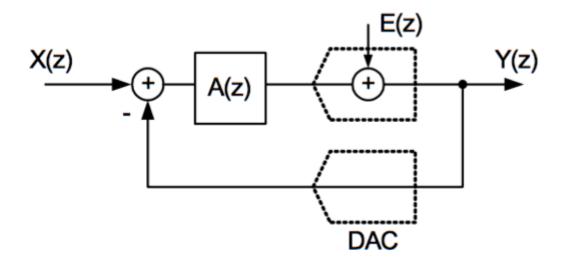
- For doubling of M we get 3dB improvement, which is the same as 1/2 a bit of accuracy
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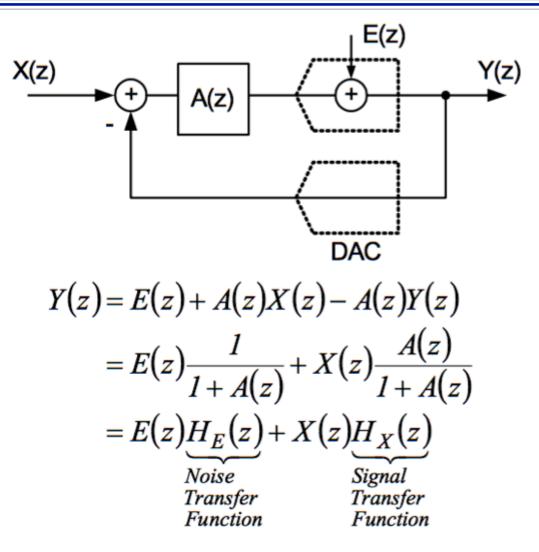


- Idea: "Somehow" build an ADC that has most of its quantization noise at high frequencies
- □ Key: Feedback

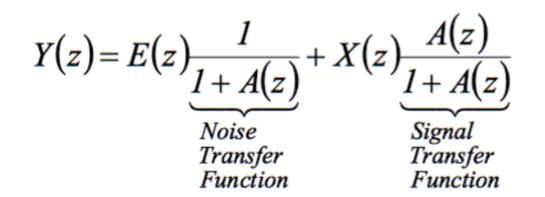






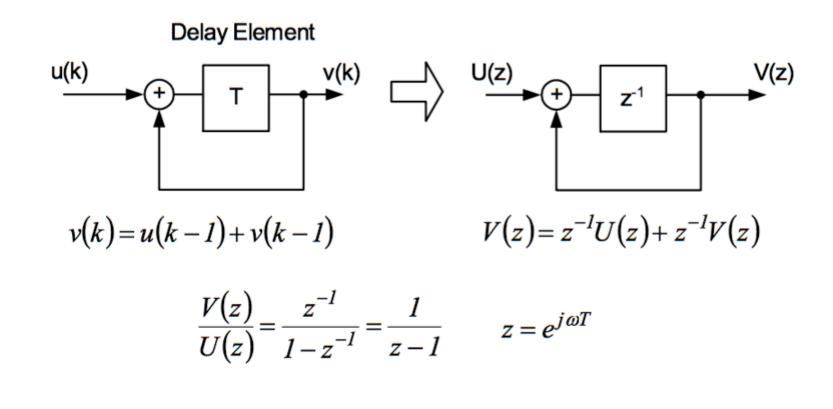


Noise Shaping Using Feedback



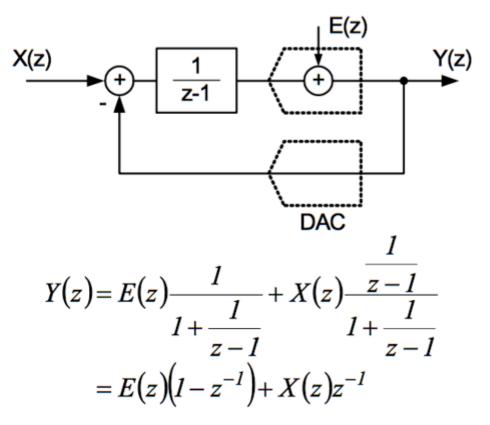
- Objective
  - Want to make STF unity in the signal frequency band
  - Want to make NTF "small" in the signal frequency band
- □ If the frequency band of interest is around DC  $(0...f_B)$  we achieve this by making |A(z)| >> 1 at low frequencies
  - Means that NTF << 1
  - Means that STF = 1





□ "Infinite gain" at DC ( $\omega = 0, z=1$ )

First Order Sigma-Delta Modulator

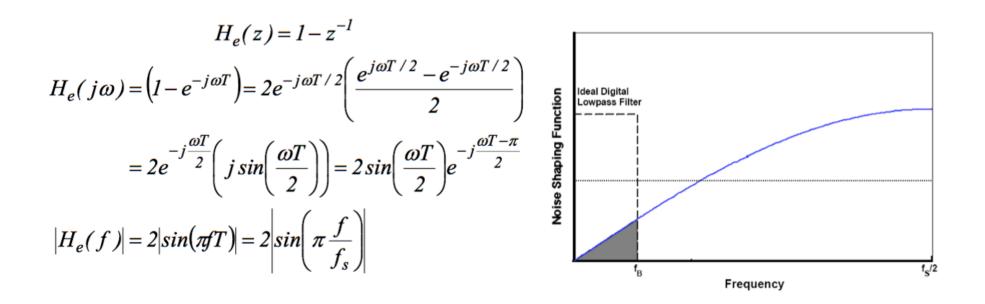


Output is equal to delayed input plus filtered quantization noise



$$H_e(z) = l - z^{-l}$$



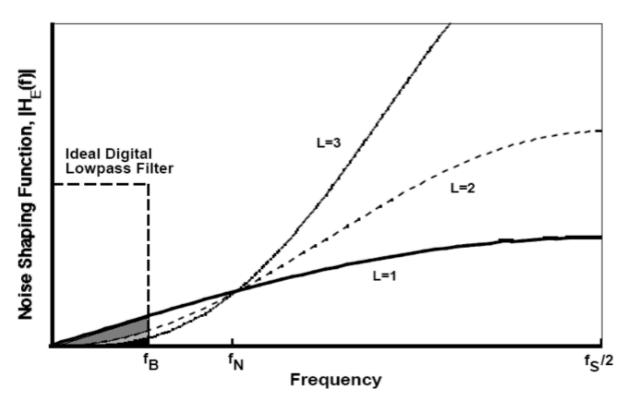


- "First order noise Shaping"
  - Quantization noise is attenuated at low frequencies, amplified at high frequencies

Higher Order Noise Shaping

□ L<sup>th</sup> order noise transfer function

$$H_E(z) = \left(l - z^{-l}\right)^L$$



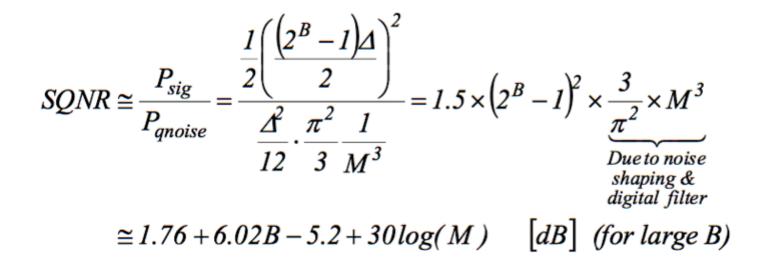
## In-Band Quantization Noise

- Question: If we had an ideal digital lowpass, what is the achieved SQNR as a function of oversampling ratio?
- Can integrate shaped quantization noise spectrum up to  $f_B$  and compare to full-scale signal

$$P_{qnoise} = \int_{0}^{f_B} \frac{\Delta^2}{12} \cdot \frac{2}{f_s} \cdot \left[ 2 \sin\left(\pi \frac{f}{f_s}\right) \right]^2 df$$
$$\approx \int_{0}^{f_B} \frac{\Delta^2}{12} \cdot \frac{2}{f_s} \cdot \left[ 2\pi \frac{f}{f_s} \right]^2 df$$
$$\approx \frac{\Delta^2}{12} \cdot \frac{\pi^2}{3} \left[ \frac{2f_B}{f_s} \right]^3 = \frac{\Delta^2}{12} \cdot \frac{\pi^2}{3} \frac{1}{M^3}$$

In-Band Quantization Noise

• Assuming a full-scale sinusoidal signal, we have



□ Each 2x increase in M results in 8x SQNR improvement

Also added <sup>1</sup>/<sub>2</sub> bit resolution



- Increasing M by 2x, means 3-dB reduction in quantization noise power, and thus 1/2 bit increase in resolution
  - "1/2 bit per octave"
- □ Is this useful?
- **•** Reality check
  - Want 16-bit ADC, f<sub>B</sub>=1MHz
  - Use oversampled 8-bit ADC with digital lowpass filter
  - 8-bit increase in resolution necessitates oversampling by 16 octaves

$$f_s \ge 2 \cdot f_B \cdot M = 2 \cdot 1MHz \cdot 2^{16}$$
$$\ge 131GHz$$

SQNR Improvement

- **D** Example Revisited
  - Want 16-bit ADC, f<sub>B</sub>=1MHz
  - Use oversampled 8-bit ADC, first order noise shaping and (ideal) digital lowpass filter
    - SQNR improvement compared to case without oversampling is -5.2dB+30log(M)
  - 8-bit increase in resolution (48 dB SQNR improvement) would necessitate M≅60 → f<sub>S</sub>=120MHz
- □ Not all that bad!

м	SQNR improvement
16	31dB (~5 bits)
256	67dB (~11 bits)
1024	85dB (~14 bits)



- Quantizers
  - Introduces quantization noise
- Data Converters
  - Oversampling to reduce interference and quantization noise → increase ENOB (effective number of bits)
  - Practical DACs use practical interpolation and reconstruction filters with oversampling
- Noise Shaping
  - Use feedback to reduce oversampling factor



- □ HW 5 due Sunday
- □ Midterm Next week Th 3/5
  - During class
    - Starts at exactly 4:30pm, ends at exactly 5:50pm (80 minutes)
  - Location DRLB A2
  - Old exams posted on previous years' websites
    - Disclaimer: exams before 2020 covered more material
  - Covers Lec 1- 11
  - Closed book, one (8.5x11 front/back) page cheat sheet allowed
  - Calculators allowed, no smart phones