

ESE 531: Digital Signal Processing

Lec 11: February 16th, 2017
Data Converters, Noise Shaping



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Lecture Outline

- Multi-Rate Filter Banks (con't)
- Data Converters
 - Anti-aliasing
 - ADC
 - Quantization
 - Practical DAC
- Noise Shaping

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Multi-Rate Filter Banks

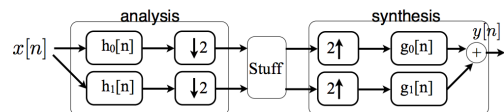
- Use filter banks to operate on a signal differently in different frequency bands
 - To save computation, reduce the rate after filtering

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Multi-Rate Filter Banks

- Use filter banks to operate on a signal differently in different frequency bands
 - To save computation, reduce the rate after filtering
- $h_0[n]$ is low-pass, $h_1[n]$ is high-pass
 - Often $h_1[n] = e^{j\pi n} h_0[n] \leftarrow$ shift freq resp by π

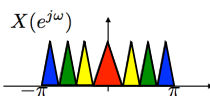
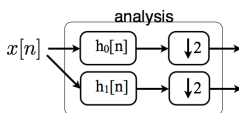


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

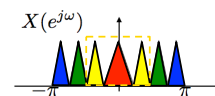
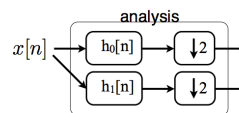


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Multi-Rate Filter Banks

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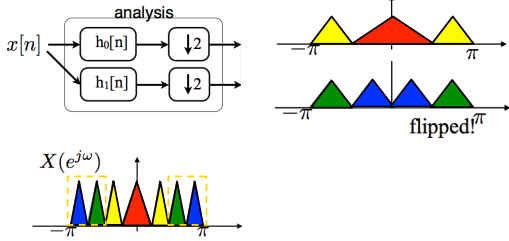


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Multi-Rate Filter Banks

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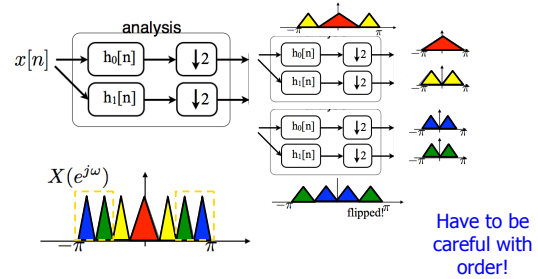


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Multi-Rate Filter Banks

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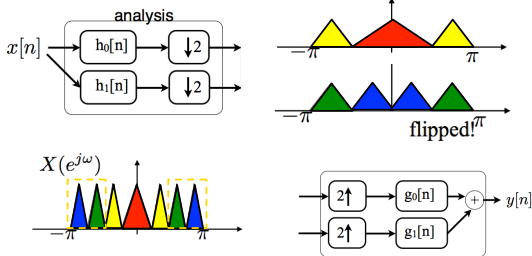


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Multi-Rate Filter Banks

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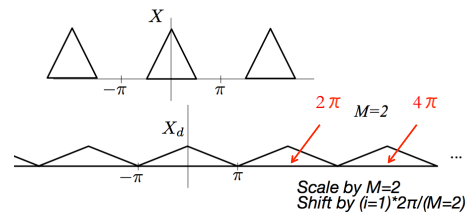


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Downsampling Reminder: Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X\left(e^{j\left(\frac{\omega}{M} - \frac{2\pi}{M}i\right)}\right)$$

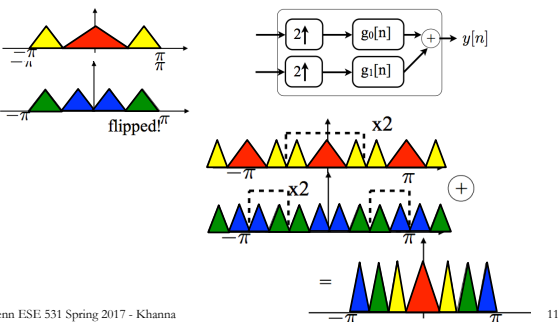


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

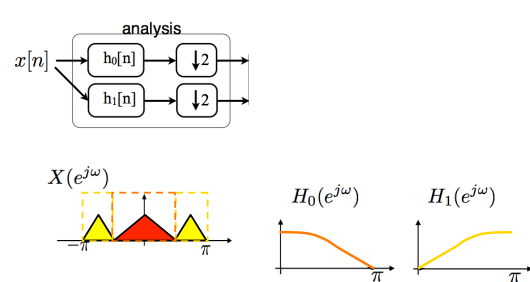


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Multi-Rate Filter Banks

- h_0, h_1 are NOT ideal low/high pass

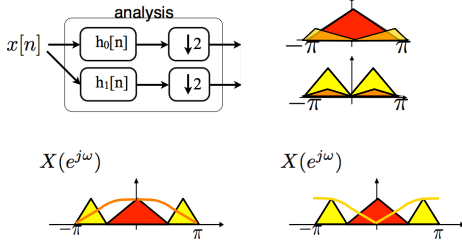


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Non Ideal Filters

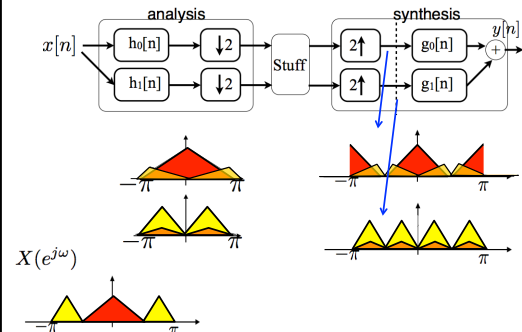
□ h_0, h_1 are NOT ideal low/high pass



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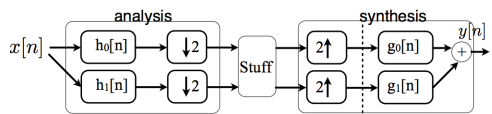
Non Ideal Filters



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Perfect Reconstruction non-Ideal Filters



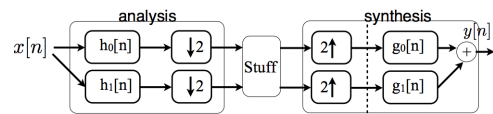
$$Y(e^{j\omega}) = \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j\omega}) + G_1(e^{j\omega})H_1(e^{j\omega})] X(e^{j\omega}) + \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j(\omega-\pi)}) + G_1(e^{j\omega})H_1(e^{j(\omega-\pi)})] X(e^{j(\omega-\pi)})$$

need to cancel! aliasing

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Quadrature Mirror Filters



Quadrature mirror filters

$$\begin{aligned} H_1(e^{j\omega}) &= H_0(e^{j(\omega-\pi)}) \\ G_0(e^{j\omega}) &= 2H_0(e^{j\omega}) \\ G_1(e^{j\omega}) &= -2H_1(e^{j\omega}) \end{aligned}$$

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Perfect Reconstruction non-Ideal Filters

$$Y(e^{j\omega}) = \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j\omega}) + G_1(e^{j\omega})H_1(e^{j\omega})] X(e^{j\omega}) + \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j(\omega-\pi)}) + G_1(e^{j\omega})H_1(e^{j(\omega-\pi)})] X(e^{j(\omega-\pi)})$$

need to cancel! aliasing

$$\begin{aligned} H_1(e^{j\omega}) &= H_0(e^{j(\omega-\pi)}) \\ G_0(e^{j\omega}) &= 2H_0(e^{j\omega}) \\ G_1(e^{j\omega}) &= -2H_1(e^{j\omega}) \end{aligned}$$

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ADC



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Aliasing

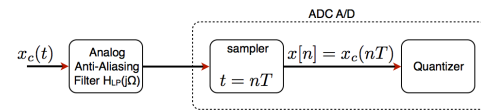
- If $\Omega_N > \Omega_s/2$, $x_r(t)$ an aliased version of $x_c(t)$

$$X_r(j\Omega) = \begin{cases} TX_s(j\Omega) & \text{if } |\Omega| \leq \Omega_s/2 \\ 0 & \text{otherwise} \end{cases}$$

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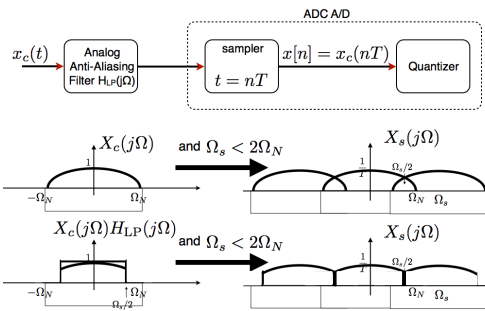
Anti-Aliasing Filter with ADC



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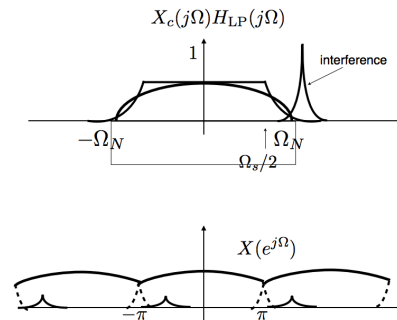
Anti-Aliasing Filter with ADC



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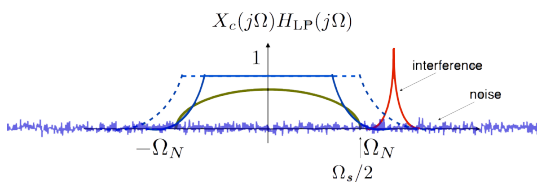
Non-Ideal Anti-Aliasing Filter



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Non-Ideal Anti-Aliasing Filter

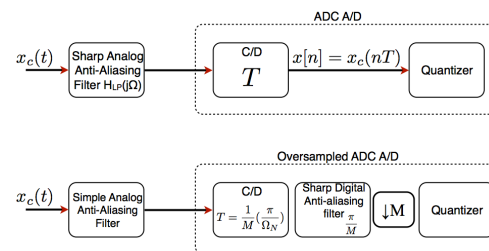


- Problem: Hard to implement sharp analog filter
- Solution: Crop part of the signal and suffer from noise and interference

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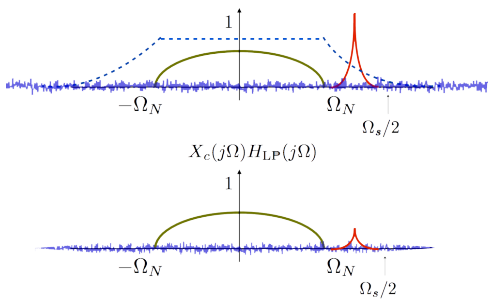
Oversampled ADC



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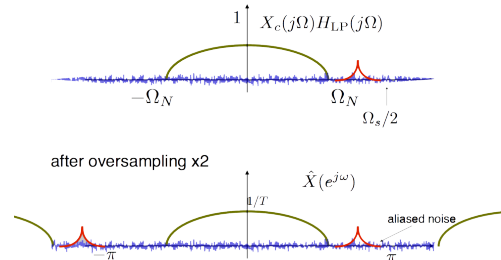
Oversampled ADC



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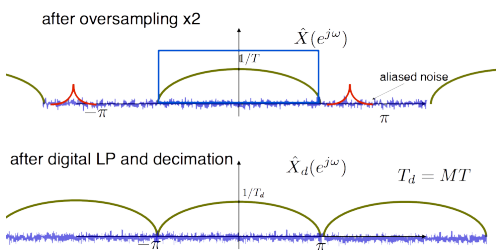
Oversampled ADC



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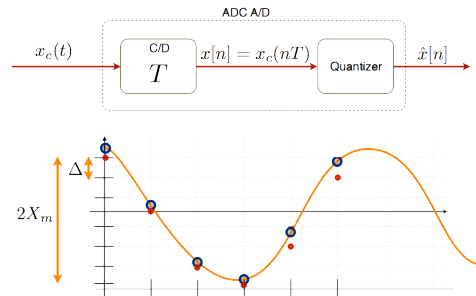
Oversampled ADC



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Sampling and Quantization



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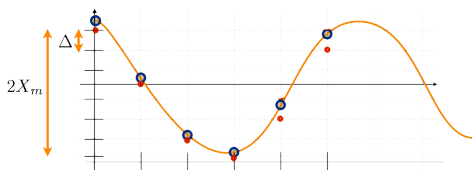
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Sampling and Quantization

for 2's complement with B+1 bits $-1 \leq \hat{x}_B[n] < 1$

$$\Delta = \frac{2X_m}{2^{B+1}} = \frac{X_m}{2^B}$$

$$\hat{x}[n] = X_m \hat{x}_B[n]$$



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Quantization Error



Model quantization error as noise

$$x[n] \rightarrow \hat{x}[n] = x[n] + e[n]$$

In that case:

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

$$(-X_m - \Delta/2) < x[n] \leq (X_m - \Delta/2)$$

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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?

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Quantization Error Statistics

- Crude assumption: $e_q(x)$ 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

$p(e_q)$

e_q

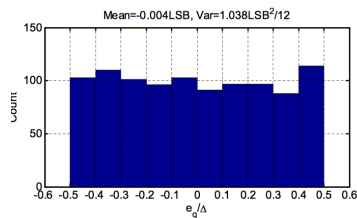
Mean $\bar{e} = \int_{-\Delta/2}^{+\Delta/2} e \cdot \frac{1}{\Delta} de = 0$

Variance $\bar{e^2} = \int_{-\Delta/2}^{+\Delta/2} e^2 \cdot \frac{1}{\Delta} de = \frac{\Delta^2}{12}$

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Reality Check

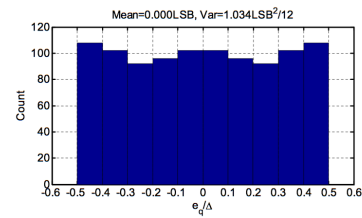


- Shown below is a histogram of e_q in an 8-bit quantizer
 - Input sequence consists of 1000 samples with Gaussian distribution, $4\sigma = \text{FSR}$

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Reality Check

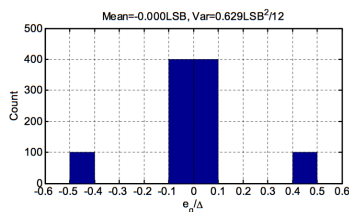


- Same as before, but now using a sinusoidal input signal with $f_{sig}/f_s = 101/1000$

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Reality Check



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

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Analysis

$$v_{sig}(n) = \cos\left(2\pi \cdot \frac{f_{sig}}{f_s} \cdot n\right)$$

- 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 of at best of 10 different values, even though we took 1000 samples

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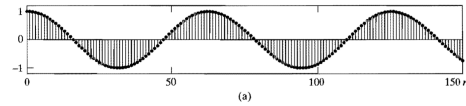
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Noise Model for Quantization Error

- 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 $m \neq n$ (white noise)
 - $e[n] \sim U[-\Delta/2, \Delta/2]$ (uniform pdf)
- Result:
- Variance is: $\sigma_e^2 = \frac{\Delta^2}{12}$, or $\sigma_e^2 = \frac{2^{-2B} X_m^2}{12}$ since $\Delta = 2^{-B} X_m$
- Assumptions work well for signals that change rapidly, are not clipped, and for small Δ

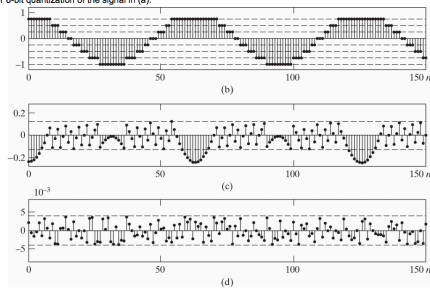
Quantization Noise

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



Quantization Noise

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).



Signal-to-Quantization-Noise Ratio

- For uniform B+1 bits quantizer

$$\begin{aligned} SNR_Q &= 10 \log_{10} \left(\frac{\sigma_x^2}{\sigma_e^2} \right) \\ &= 10 \log_{10} \left(\frac{12 \cdot 2^{2B} \sigma_x^2}{X_m^2} \right) \end{aligned}$$

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

Signal-to-Quantization-Noise Ratio

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

- 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)

Signal-to-Quantization-Noise Ratio

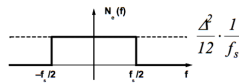
- Assuming full-scale sinusoidal input, we have

$$SQNR = \frac{P_{\text{sig}}}{P_{\text{noise}}} = \frac{1}{2} \left(\frac{2^B \Delta}{2} \right)^2 \frac{12}{\Delta^2} = 1.5 \times 2^{2B} = 6.02B + 1.76 \text{ dB}$$

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



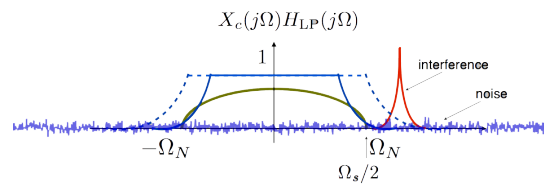
References

- 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.

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Non-Ideal Anti-Aliasing Filter

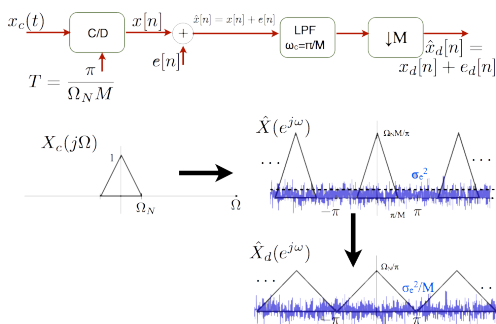


- Problem: Hard to implement sharp analog filter
- Solution: Crop part of the signal and suffer from noise and interference

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Quantization Noise with Oversampling



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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

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

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

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Practical DAC

Practical DAC

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

- Scaled train of sinc pulses
- Difficult to generate sinc \rightarrow Too long!

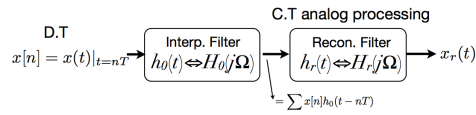


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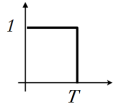
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Practical DAC



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



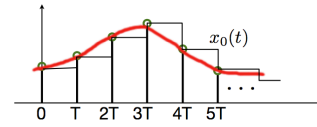
$$H_0(j\Omega) = T e^{-j\Omega \frac{T}{2}} \text{sinc}\left(\frac{\Omega}{\Omega_s}\right)$$

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Practical DAC

Zero-Order-Hold interpolation



$$x_0(t) = \sum_{n=-\infty}^{\infty} x[n]h_0(t - nT) = h_0(t) * x_s(t)$$

Taking a FT:

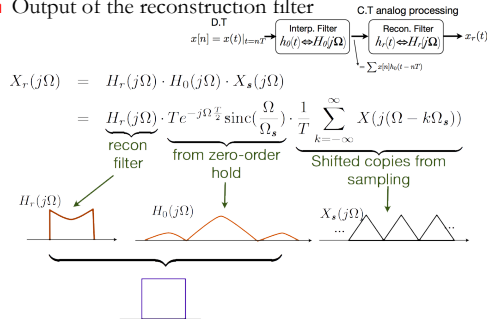
$$\begin{aligned} X(j\Omega) &= H_0(j\Omega)X_s(j\Omega) \\ &= H_0(j\Omega) \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\Omega - k\Omega_s)) \end{aligned}$$

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Practical DAC

- Output of the reconstruction filter



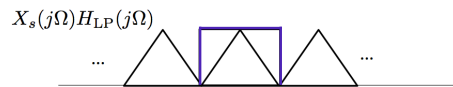
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Practical DAC



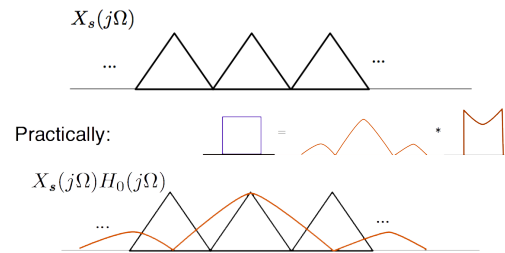
Ideally:



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Practical DAC



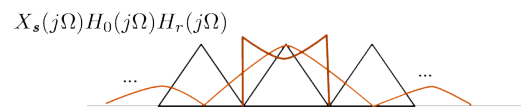
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Practical DAC



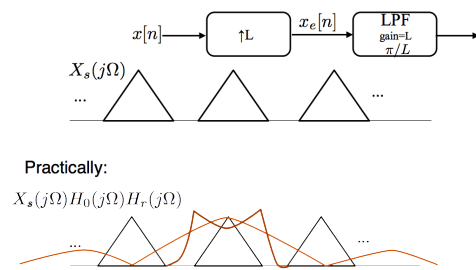
Practically:



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Practical DAC with Upsampling



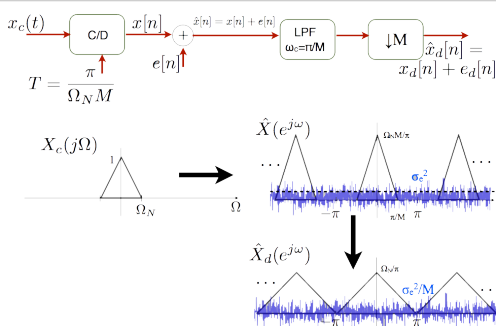
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Noise Shaping



Quantization Noise with Oversampling



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Quantization Noise with Oversampling

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 - No filtering of signal!
- Noise variance is reduced by factor of M

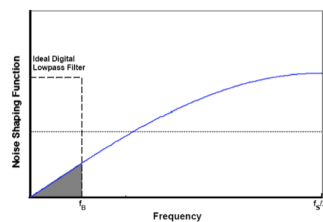
$$\text{SNR}_Q = 6.02B + 10.8 - 20 \log_{10} \left(\frac{X_m}{\sigma_x} \right) + \boxed{10 \log_{10} M}$$

- 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!

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Noise Shaping

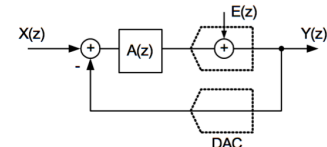


- Idea: "Somehow" build an ADC that has most of its quantization noise at high frequencies
- Key: Feedback

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Noise Shaping Using Feedback



$$\begin{aligned} Y(z) &= E(z) + A(z)X(z) - A(z)Y(z) \\ &= E(z) \frac{1}{1 + A(z)} + X(z) \frac{A(z)}{1 + A(z)} \\ &= E(z) \underbrace{\frac{1}{1 + A(z)}}_{\text{Noise Transfer Function}} + X(z) \underbrace{\frac{A(z)}{1 + A(z)}}_{\text{Signal Transfer Function}} \end{aligned}$$

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Noise Shaping Using Feedback

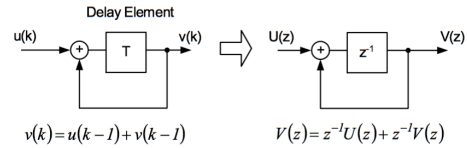
$$Y(z) = E(z) \underbrace{\frac{1}{1+A(z)}}_{\text{Noise Transfer Function}} + X(z) \underbrace{\frac{A(z)}{1+A(z)}}_{\text{Signal Transfer Function}}$$

- 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 ($0 \dots f_B$) we achieve this by making $|A(z)| \gg 1$ at low frequencies
 - Means that NTF $\ll 1$
 - Means that STF ≈ 1

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Discrete Time Integrator



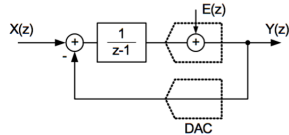
$$\frac{V(z)}{U(z)} = \frac{z^{-1}}{1-z^{-1}} = \frac{1}{z-1} \quad z = e^{j\omega T}$$

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

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First Order Sigma-Delta Modulator



$$Y(z) = E(z) \frac{1}{1+\frac{1}{z-1}} + X(z) \frac{\frac{1}{z-1}}{1+\frac{1}{z-1}}$$

$$= E(z) \left(\frac{1}{1-z^{-1}} \right) + X(z) z^{-1}$$

- Output is equal to delayed input plus filtered quantization noise

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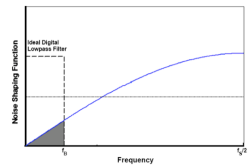
NTF Frequency Domain Analysis

$$H_e(z) = 1 - z^{-1}$$

$$H_e(j\omega) = (1 - e^{-j\omega T}) = 2e^{-j\omega T/2} \left(\frac{e^{j\omega T/2} - e^{-j\omega T/2}}{2} \right)$$

$$= 2e^{-j\omega T/2} \left(j \sin\left(\frac{\omega T}{2}\right) \right) = 2 \sin\left(\frac{\omega T}{2}\right) e^{-j\omega T/2}$$

$$|H_e(f)| = 2 \sin(\pi f T) = 2 \sin\left(\pi \frac{f}{f_s}\right)$$



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

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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

$$P_{\text{noise}} = \int_0^{f_B} \frac{\Delta^2}{12} \cdot \frac{2}{f_s} \cdot \left[2 \sin\left(\pi \frac{f}{f_s}\right) \right]^2 df$$

$$\approx \int_0^{f_B} \frac{\Delta^2}{12} \cdot \frac{2}{f_s} \cdot \left[2\pi \frac{f}{f_s} \right]^2 df$$

$$\approx \frac{\Delta^2}{12} \cdot \frac{\pi^2}{3} \left[\frac{2f_B}{f_s} \right]^3 = \frac{\Delta^2}{12} \cdot \frac{\pi^2}{3} \frac{1}{M^3}$$

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In-Band Quantization Noise

- Assuming a full-scale sinusoidal signal, we have

$$SQNR \approx \frac{P_{\text{sig}}}{P_{\text{noise}}} = \frac{\frac{1}{2} \left(\frac{2^B - 1}{2} \Delta \right)^2}{\frac{\Delta^2}{12} \cdot \frac{\pi^2}{3} \frac{1}{M^3}} = 1.5 \times (2^B - 1)^2 \times \frac{3}{\pi^2} \times M^3$$

Due to noise shaping & digital filter

$$\approx 1.76 + 6.02B - 5.2 + 30 \log(M) \quad [\text{dB}] \quad (\text{for large } B)$$

- Each 2x increase in M results in 8x SQNR improvement
 - Also added 1/2 bit resolution

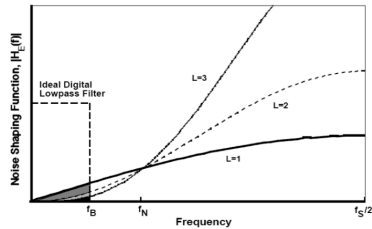
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Higher Order Noise Shaping

- L^{th} order noise transfer function

$$H_E(z) = (1 - z^{-1})^L$$



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Big Ideas

- Multi-Rate Filter Banks (con't)
 - Operating on different frequency bands at lower sampling rates
- 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

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Admin

- HW 4 extended to Tuesday at midnight
 - Typo in code in MATLAB problem, corrected handout
 - See Piazza for more information
- New tentative HW schedule posted

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