

Fixed Point Effects in Digital Filters

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Finite-precision Problems

- Quantizers are nonlinear devices
 - Characteristics may be significantly different from the ideal linear filter
- Overflow
- Coefficient quantization
- Limit-Cycle Oscillations

Quantizers

- Nonlinear effects make it extremely difficult to precisely analyze the filter's performance.
- How do we model a fixed-point filter then?
 - Adopt a statistical model of the quantization effects
 - Results in a linear model for the filter

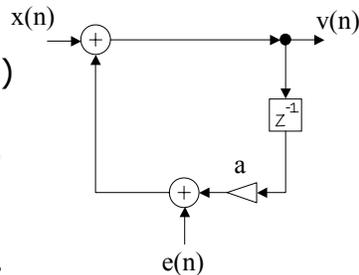
Statistical Characterization

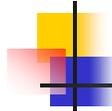
$$v(n) = Q[av(n-1)] + x(n)$$

$$Q[av(n-1)] = av(n-1) + e(n)$$

$$v(n) = av(n-1) + x(n) + e(n)$$

- We can now view the response of the filter as coming from two inputs.





Basic Assumptions

- The noise source is stationary white noise.
 - The sequence $e(n)$ is uncorrelated with the sequence $e(m)$ for $n \neq m$.
 - Sequence is mean ergodic and correlation ergodic.
- The error sequence $e(n)$ is uncorrelated with the input sequence $x(n)$.



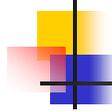
Mean

- The mean of the output generated by the filter with impulse response $h(n)$ when excited by $e(n)$ is

$$m_q = m_e \sum_{n=-\infty}^{\infty} h(n)$$

equivalently put

$$m_q = m_e H(0)$$



Autocorrelation

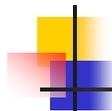
The autocorrelation is computed to be

$$\gamma_{qq}(n) = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} h(k)h(l)\gamma_{ee}(k-l+n)$$

This reduces to

$$\sigma_q^2 = \sigma_e^2 \sum_{k=-\infty}^{\infty} h^2(k)$$

By Parseval's theorem $\sigma_q^2 = \frac{\sigma_e^2}{2\pi} \int_{-\pi}^{\pi} |H(\omega)|^2 d\omega$



Types of Quantization

b fractional bits

- Rounding $m_e = 0$ $\sigma_e^2 = \frac{2^{-2b}}{12}$
- Truncation $m_e = -\frac{2^{-b}}{2}$ $\sigma_e^2 = \frac{2^{-2b}}{12}$
- Magnitude Truncation $m_e = 0$ $\sigma_e^2 = \frac{2^{-2b}}{3}$

Section Ordering is Important

- Example

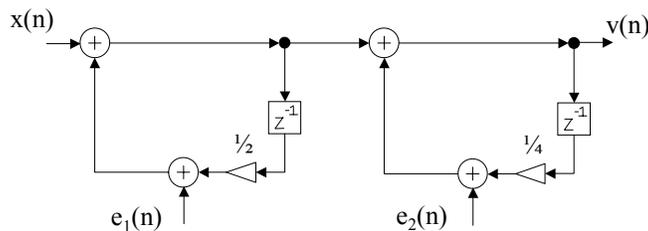
$$H(z) = H_1(z)H_2(z)$$

Where

$$H_1(z) = \frac{1}{1 - \frac{1}{2}z^{-1}} \quad H_2(z) = \frac{1}{1 - \frac{1}{4}z^{-1}}$$

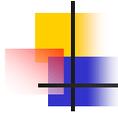
and their corresponding impulse responses are given by $h(n)$, $h_1(n)$ and $h_2(n)$

Realization 1

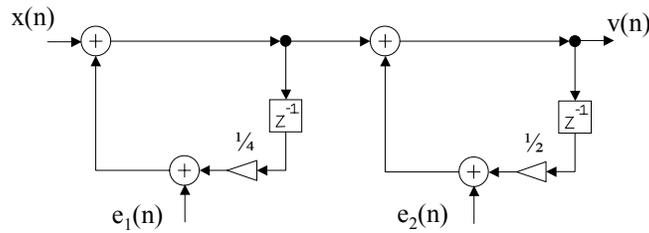


With rounding

$$\sigma_{q1}^2 = \sigma_e^2 \left[\sum_{n=0}^{\infty} h^2(n) + \sum_{n=0}^{\infty} h_2^2(n) \right] \approx 2.90\sigma_e^2$$



Realization 2



With rounding

$$\sigma_{q2}^2 = \sigma_e^2 \left[\sum_{n=0}^{\infty} h^2(n) + \sum_{n=0}^{\infty} h_1^2(n) \right] \approx 3.16\sigma_e^2$$



Section Ordering Comparison

- The overall noises were found to be

$$\begin{aligned} \sigma_{q1}^2 &\approx 2.90\sigma_e^2 & \frac{\sigma_{q2}^2}{\sigma_{q1}^2} &\approx 1.09 \\ \sigma_{q2}^2 &\approx 3.16\sigma_e^2 & & \end{aligned}$$

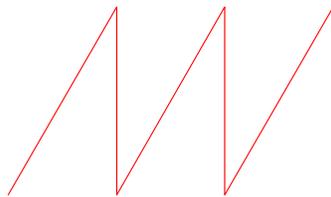
- Thus, the second realization leads to 9% more noise power than the first.

Overflow

- Wrap Around

- Ex.

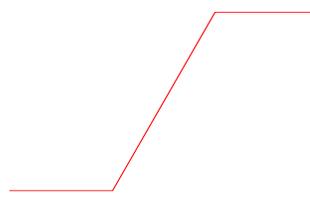
$$7 \text{ (0111)} + 1 \text{ (0001)} = \\ -8 \text{ (1000)}$$



- Saturation

- Ex.

$$5 \text{ (0101)} + 4 \text{ (0100)} = \\ 7 \text{ (0111)}$$



Scaling to Prevent Overflow

- Pessimistic Scaling

- Narrowband Scaling

$$A_x < \frac{1}{\max_{m=-\infty}^{\infty} |h_k(m)|}$$

$$A_x < \frac{1}{\max_{0 \leq \omega \leq 2\pi} |H_k(\omega)|}$$



Practical Round-Off Effects on Digital Filters

- Coefficient Quantization
 - Frequency Response Characteristics
- Internal Wordlength Quantization
 - Dynamic Range
 - Signal-to-Noise Ratio
 - Limit Cycles



Coefficient Quantization

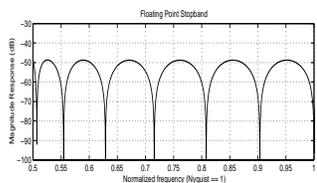
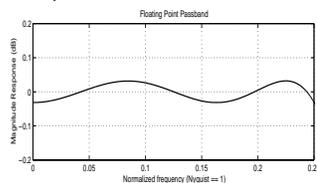
- Coefficient quantization alters the values of the coefficients => changes your frequency response
- A filter designed in floating point arithmetic to meet certain specs may not meet those specifications after coefficient quantization

Coefficient Quantization Example

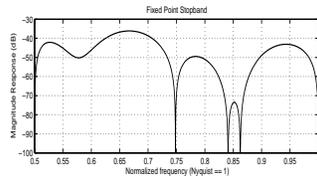
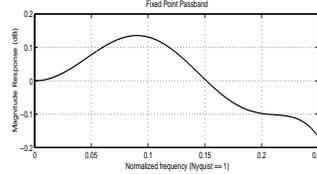
- We used the Parks-McClellan optimal FIR filter algorithm to design a 21-tap filter
- Designed the filter in floating point format
- Took each of the coefficients and rounded to the nearest 8-bit two's complement number
- Ex:
0.2011929 => .2031250
(00011010 two's complement)
- Lesson: Over-design the filter and/or use an optimization algorithm to meet the spec

Frequency Response

Floating Point
-Passband Ripple: .003 dB
-Stopband Attenuation: 48 dB



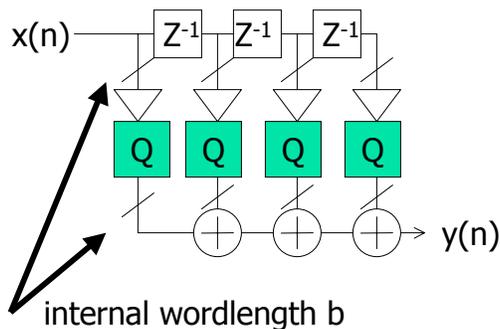
Fixed Point
-Passband Ripple: .14 dB
-Stopband Attenuation: 36 dB



Coefficient Quantization => Frequency Response Degradation

Internal Wordlength Quantization

- Quantization of internal wordlength leads to finite-wordlength effects



- 1) Dynamic Range
- 2) Signal-to-Noise Ratio
- 3) Limit Cycles

Dynamic Range Constraints

- Dynamic Range is defined as:

$$20 \log_{10} \frac{|\text{range of representable numbers}|}{|\text{smallest non-zero representable number}|}$$

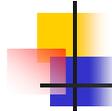
- The larger the dynamic range specification, the larger internal wordlength b required

- Ex. b -bit number $(x_0, x_1, x_2, x_3, x_4, \dots, x_{b-1})$

- DR: $20 \log |2 - 2^{-(b-1)}| / |2^{-(b-1)}|$

$$= 20 \log (2^b - 1)$$

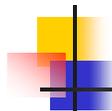
$$\sim 6 \text{ dB / bit}$$



Dynamic Range (cont'd)

- $b = 8 \Rightarrow DR \sim 48 \text{ dB}$
- $b = 12 \Rightarrow DR \sim 72 \text{ dB}$
- $b = 16 \Rightarrow DR \sim 96 \text{ dB}$

Nowadays, most hi-fi audio systems have a DR in the range of 80-100 dB



Signal-to-Noise Ratio

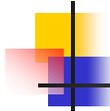
- The signal-to-noise ratio is defined as:

$$\begin{aligned} \text{SNR} &= 10 \log_{10} \text{ signal power / noise power} \\ &= 10 \log_{10} \sigma_x^2 / \sigma_e^2 \end{aligned}$$

- For our system (wordlength b):

$$\begin{aligned} \text{SNR} &= 10 \log_{10} \sigma_x^2 / (2^{-(b-1)}/12) \\ &= 10 \log_{10} \sigma_x^2 + 6.02b + 4.77 \end{aligned}$$

- Ex. $x(n) = .75 \sin(\omega n) \Rightarrow \text{SNR} = 6.02b - .739$
 $b = 8 \Rightarrow \text{SNR} = 47 \text{ dB}$



Signal-to-Noise Ratio (cont'd)

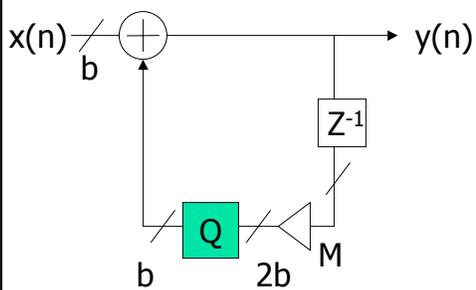
- Thus, SNR increases by ~ 6 dB / bit
- To maximize SNR, you want to scale the signal $x(n)$ as large as possible (to increase σ_x^2)
- However, there is a tradeoff—you need to keep all internal signals small enough to prevent overflow / saturation (use normalization and scaling techniques)



Limit Cycles

- Limit cycles occur when the output of a digital filter does not decay to zero when the input goes to zero
- Limit cycles occur only in IIR filters and never occur in FIR filters
- They are caused by quantizing the data after a feedback multiplier in a recursive loop

Limit-Cycle Example

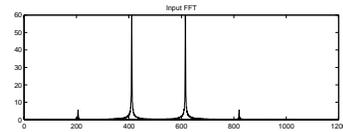
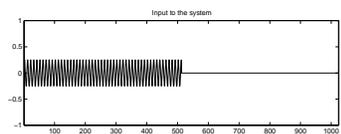


- let $M = -0.96$, $y(-1) = 14$, $x(n) = 0$ and rounding to the nearest integer

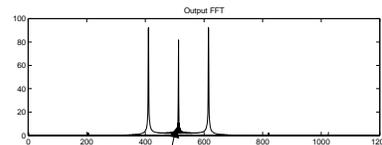
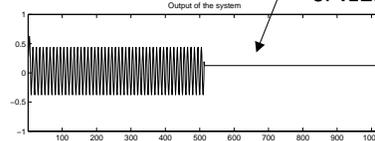
n	$-0.96 * y(n-1)$	$y(n)$
0	-13.44	-13
1	12.48	12
2	-11.52	-12
3	11.52	12

=> Limit cycle puts energy at $F_s/2$ which is detrimental

Limit Cycles in Matlab



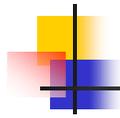
INPUT



OUTPUT

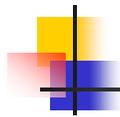
...causes a spike at DC

Limit cycle of .125 ...



Eliminating Limit Cycles

- Use magnitude truncation (which always decreases the energy of the signal)
- Use a filter for which a Lyapunov function exists
- Use controlled rounding
- Use novel filter structures designed to eliminate limit cycles



References

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- R. Schafer. *MEAD DSP IC Design Course, Lecture #3, Quantization Effects in Digital Filters*. Atlanta, Georgia.