

Analog Filter Adaptation Using a Dithered Linear Search Algorithm

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May 29, 2002



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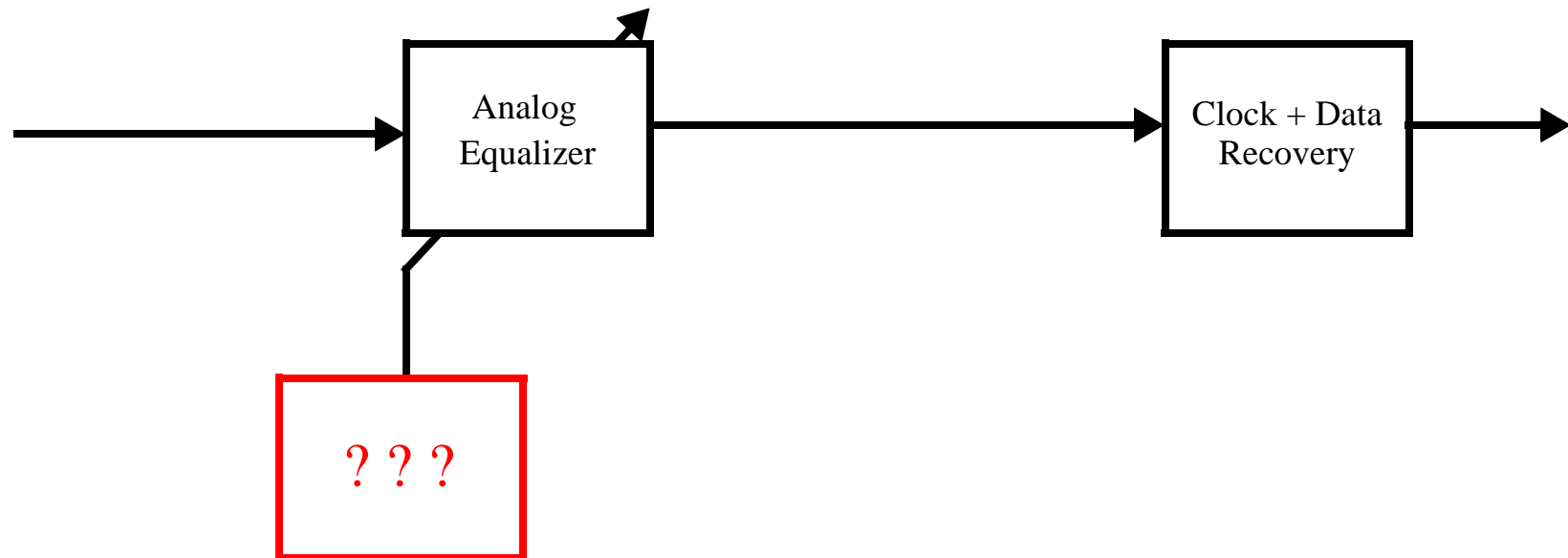
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Outline

- Motivation - analog adaptive filters
- Background - LMS algorithm
- Dithered Linear Search
- Results
- Conclusions

Motivation

- Analog adaptive filters offer several important advantages in high speed mixed signal systems:
 - reduced specifications on the A/D converter
 - reduced specifications on the analog line driver (echo cancellation application)
 - moves the equalizer outside of the timing recovery loop
 - potential for power and area savings over digital filters (at high speeds and long impulse responses)



- **Problem:** How are the filter parameters adapted/optimized???

Background - LMS Algorithm

- LMS adaptation is popular for digital filters due to its simple and robust hardware implementation
- filter parameters are updated according to:

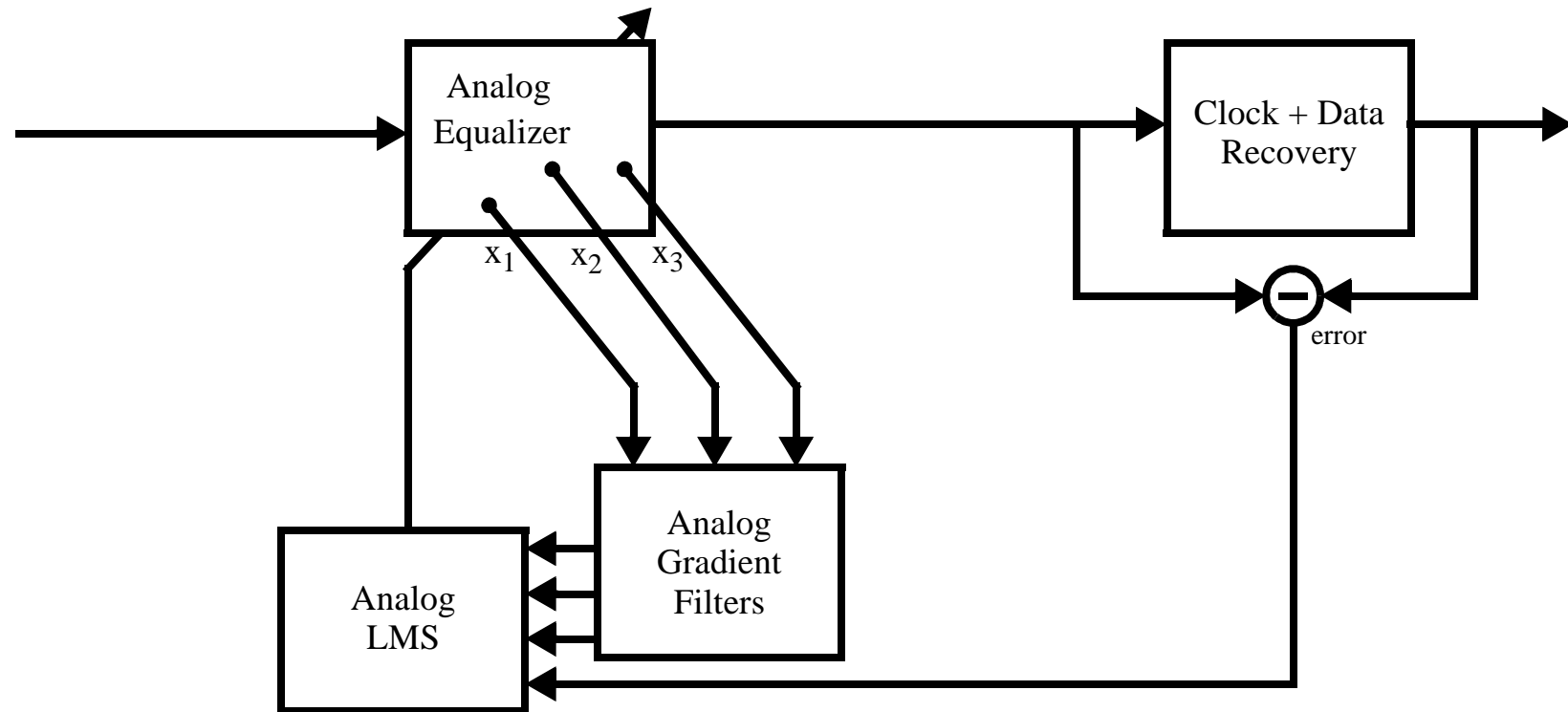
$$p(k + 1) = p(k) + 2\mu \cdot \phi(k) \cdot e(k)$$

$$\phi = \frac{\partial y}{\partial p}$$

- **Problem:** For an analog filter, the LMS algorithm is neither simple nor robust!!!
- it can be very difficult to obtain the gradient signals, ϕ
- dc offsets on the gradient and error signals lead to inaccurate convergence

Example: Equalization in Digital Communications

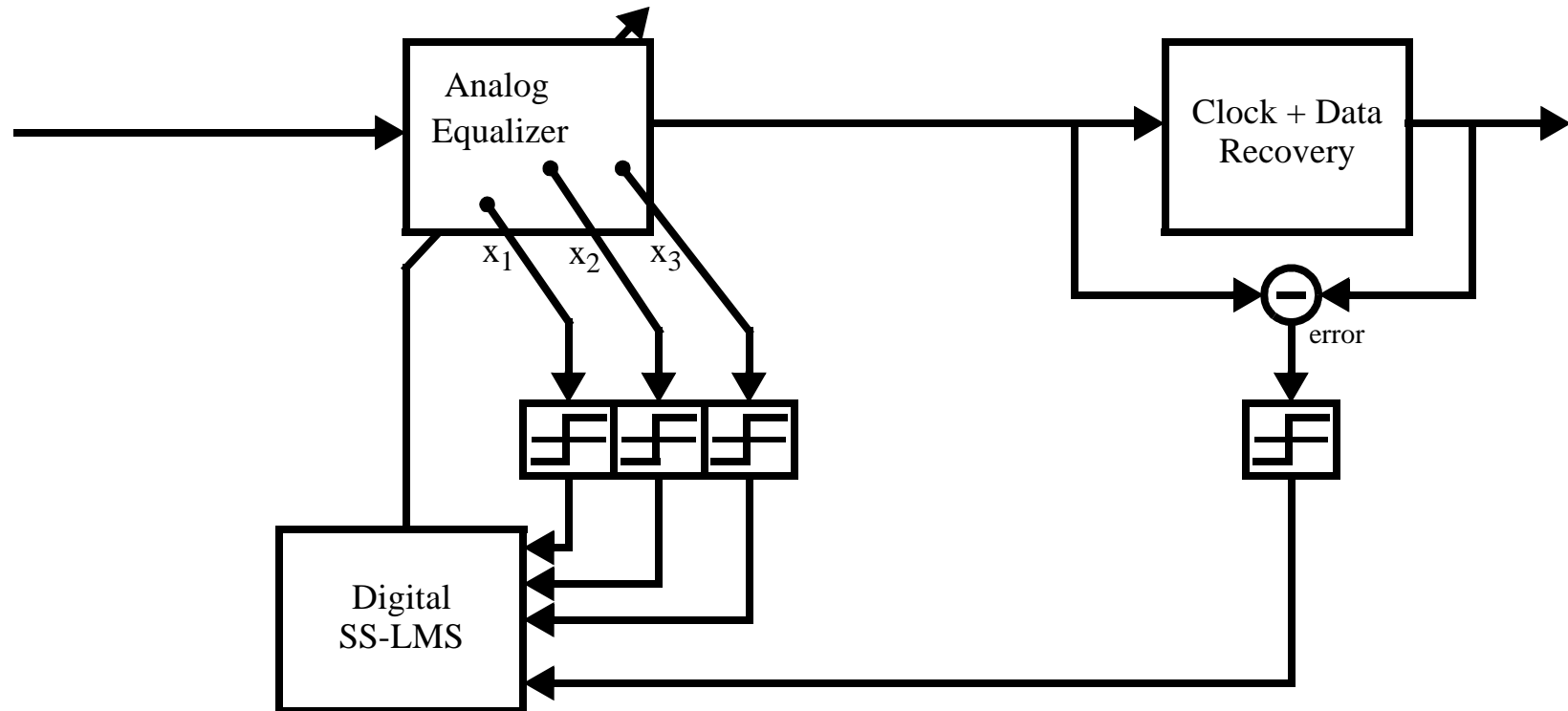
- LMS algorithm implemented with analog circuits



- lots of high-speed analog design required
- dc offsets will hinder adaptation

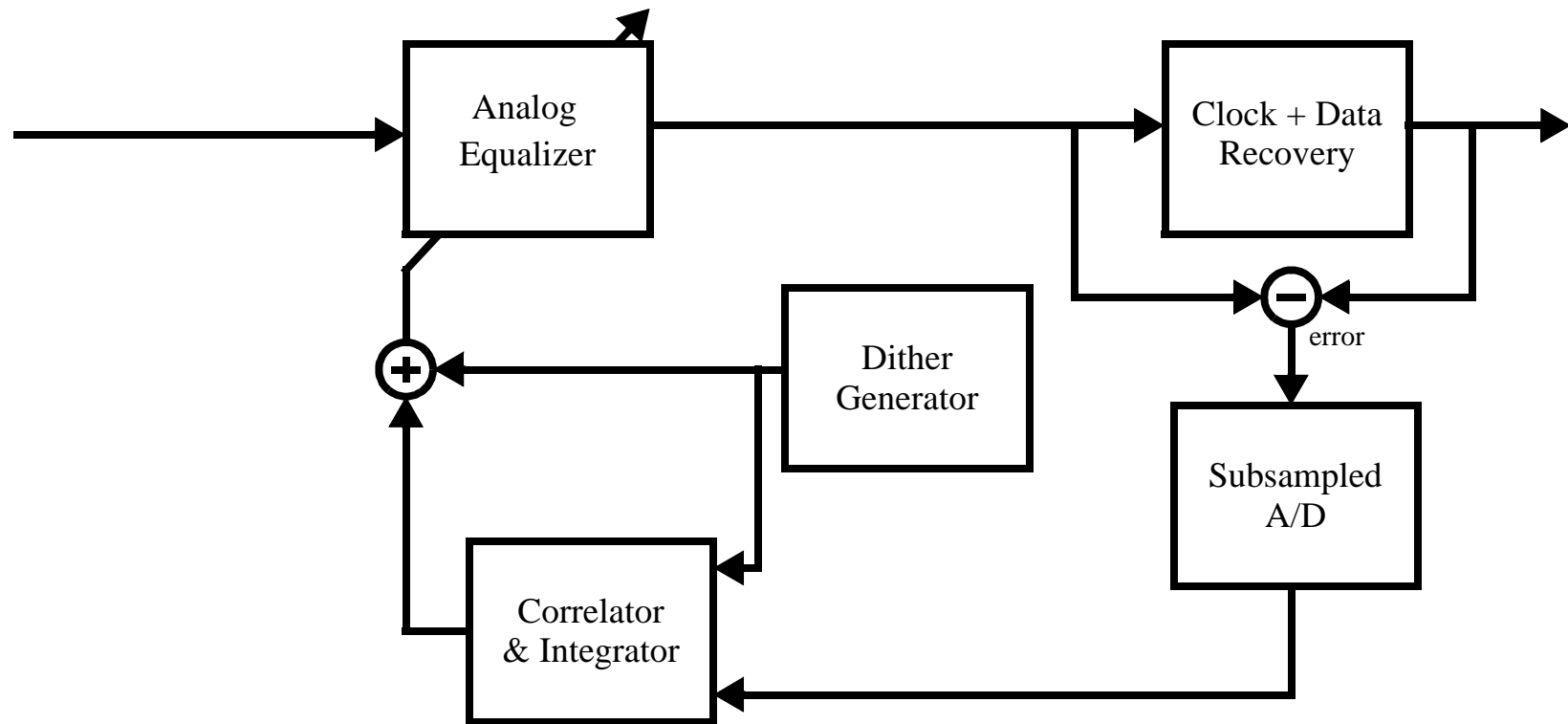
Example: Equalization in Digital Communications

- Sign-Sign LMS algorithm implemented digitally



- + comparators may be subsampled and the digital circuitry run at a slow rate
- + dc offset effects can be eliminated
- many additional comparators may be required, each loading the filter's internal nodes
- does not work for some analog equalizers (depends on the filter topology)

Dithered Linear Search



- + applicable to general filter structures
- + dc offset effects can be eliminated
- + the A/D converter and all the digital circuitry can be subsampled & hence, operated at a relatively slow rate
- adaptation can be slow due to averaging required in the correlator

Dithered Linear Search

- general gradient descent algorithm:

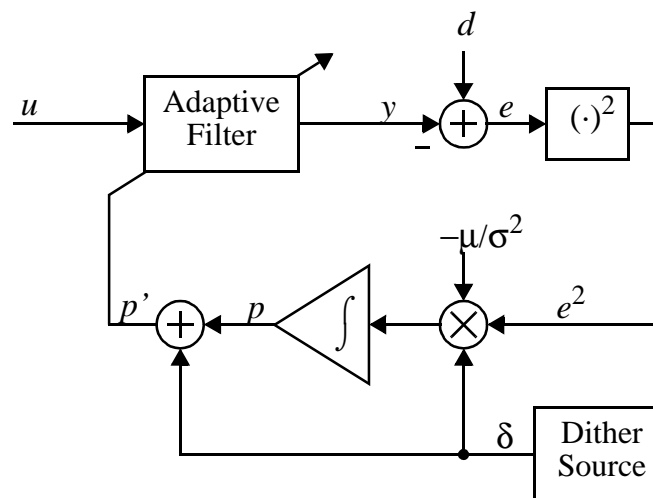
$$p(k+1) = p(k) - \mu \cdot \frac{\partial E[e^2(k)]}{\partial p(k)}$$

- can be shown that,

$$E\left[\frac{1}{\sigma^2} \cdot \delta(k) \cdot e^2(k)\right] = \frac{\partial E[e^2(k)]}{\partial p(k)}$$

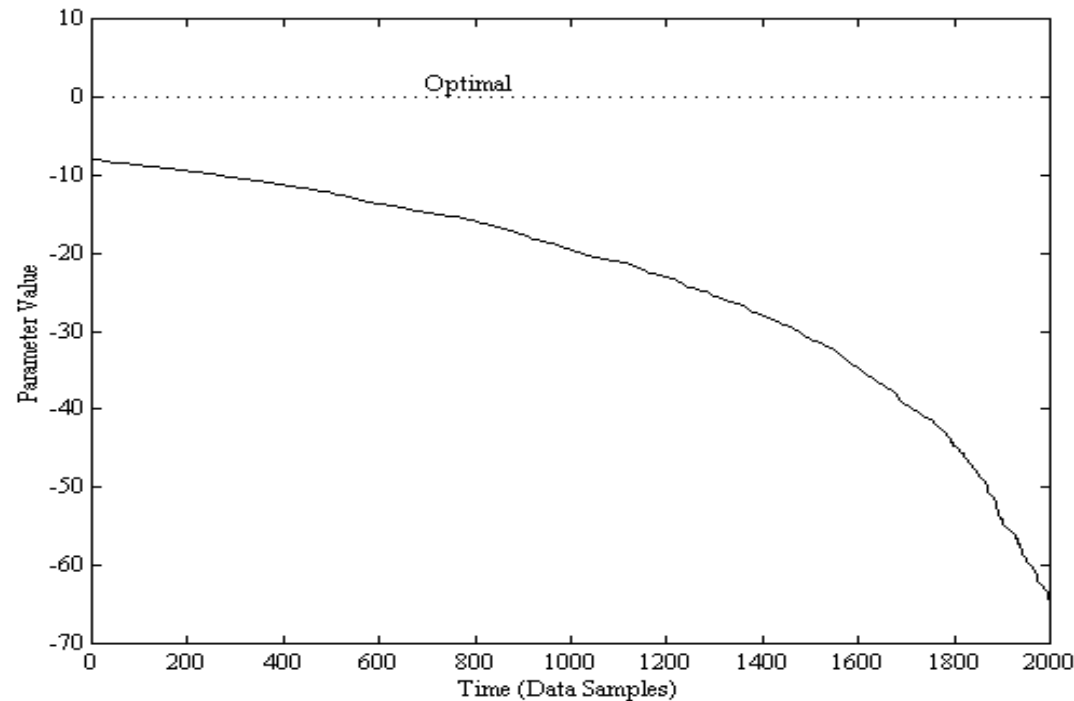
- therefore, the DLS algorithm updates the filter parameters according to:

$$p(k+1) = p(k) - \frac{\mu}{\sigma^2} \cdot \delta(k) \cdot e^2(k)$$



Choice of Dither Signal

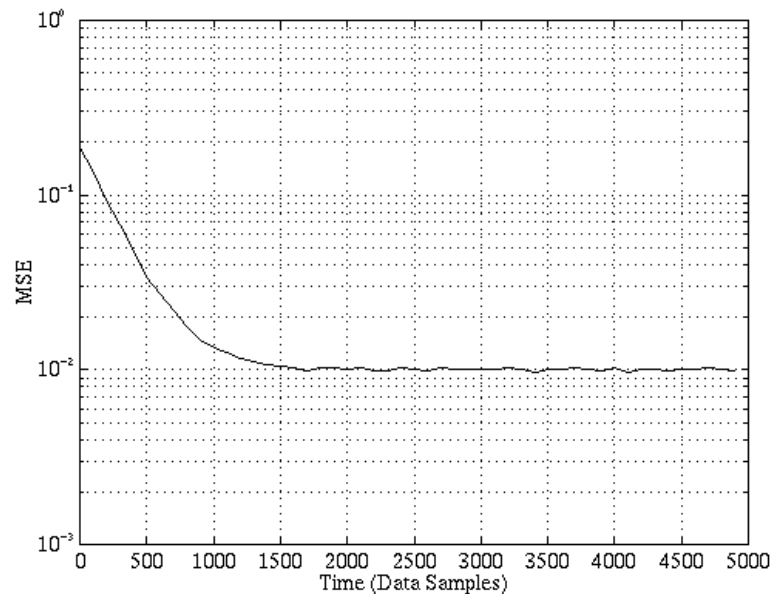
- Binary dither provides straightforward implementation
 1. pseudorandom binary sequences
 - + simple hardware implementation
 - + error introduced by the dither is random
 - long sequences of ones or zeros are possible, which can cause the algorithm to diverge



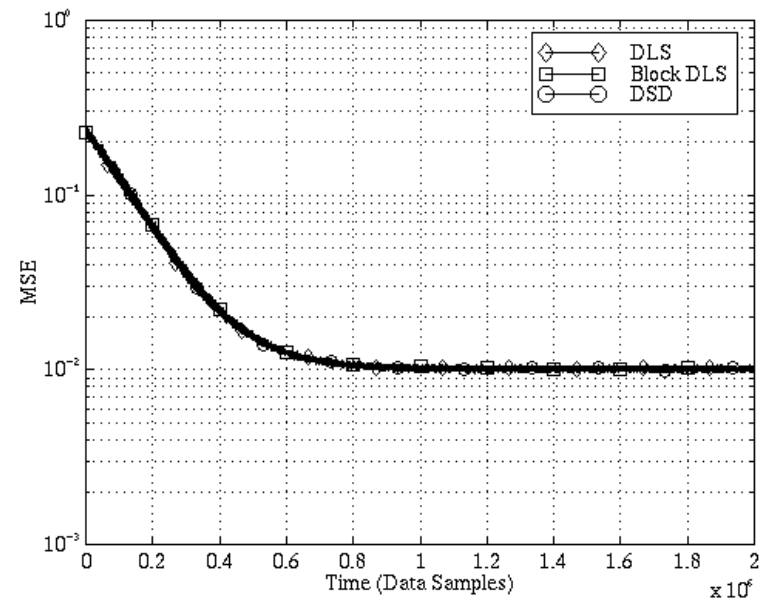
Choice of Dither Signal

- 2. orthogonal periodic binary sequences (e.g. Hadamard sequences)
 - + simple hardware implementation
 - + no long sequences of ones or zeros
 - periodic nature of the dither can cause spurs to appear in the filter output

Not a problem when dither is applied slowly, but DLS is already slow:



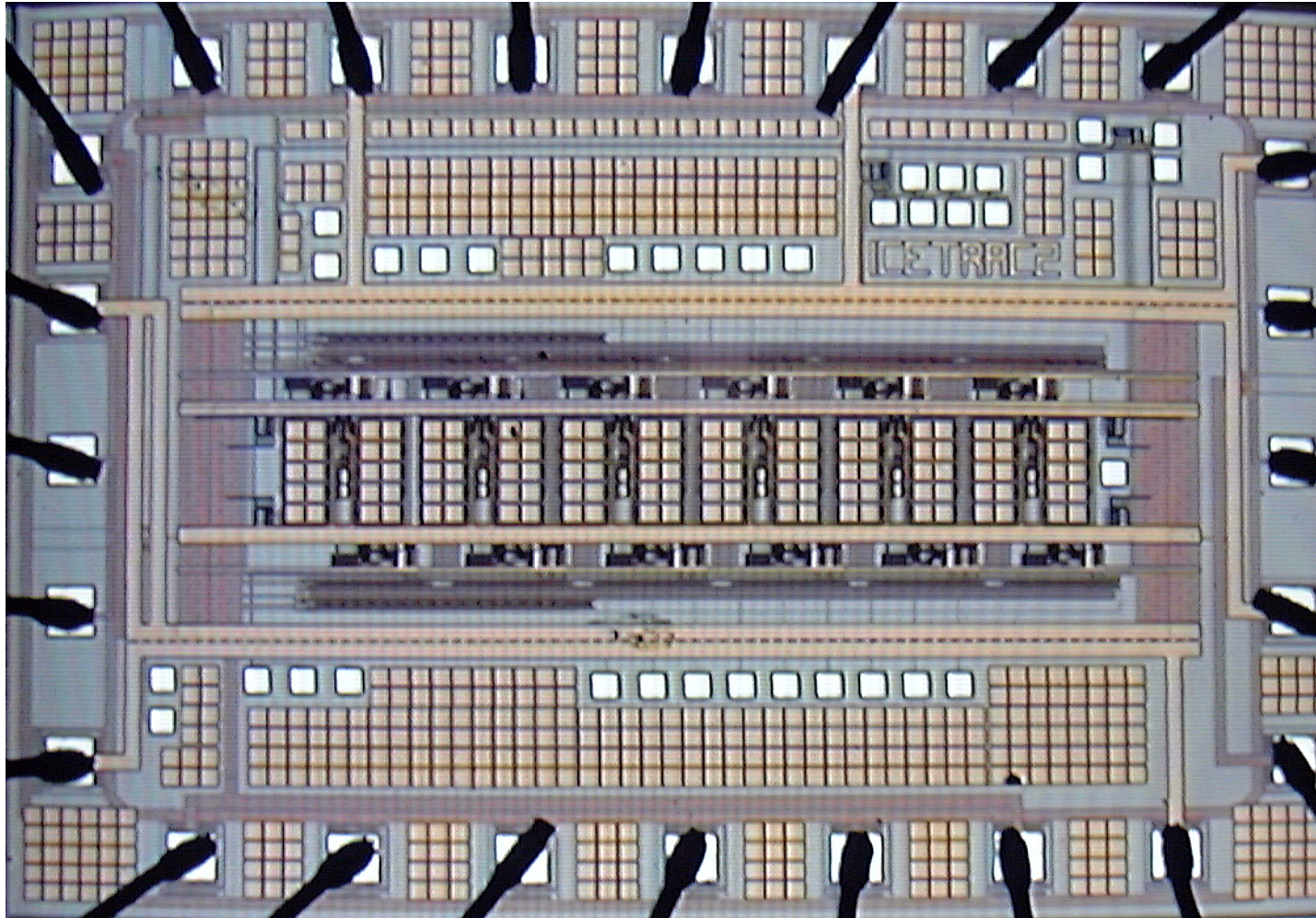
LMS



DLS

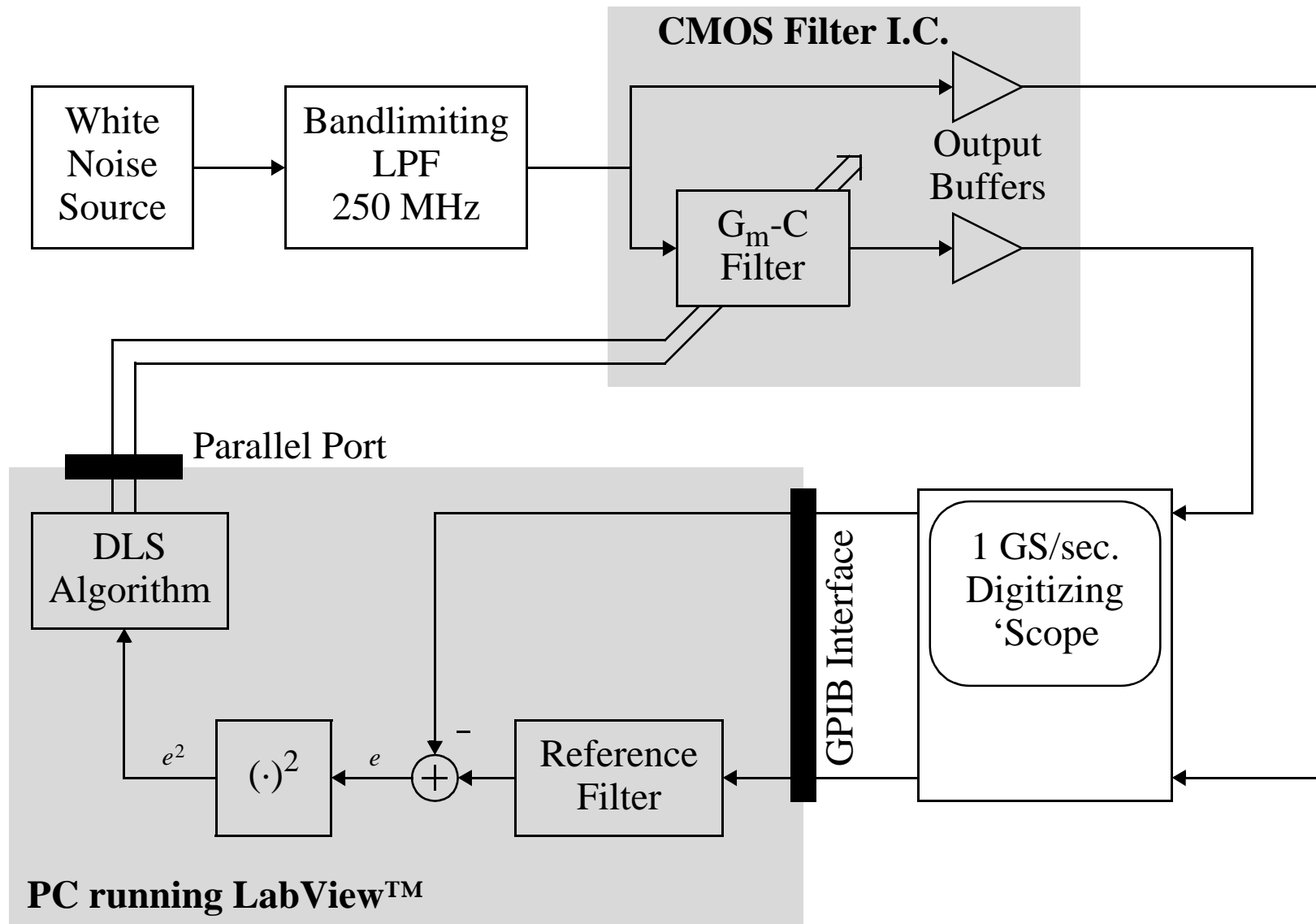
- 1000x slower for a 5-tap FIR filter

Die Photo

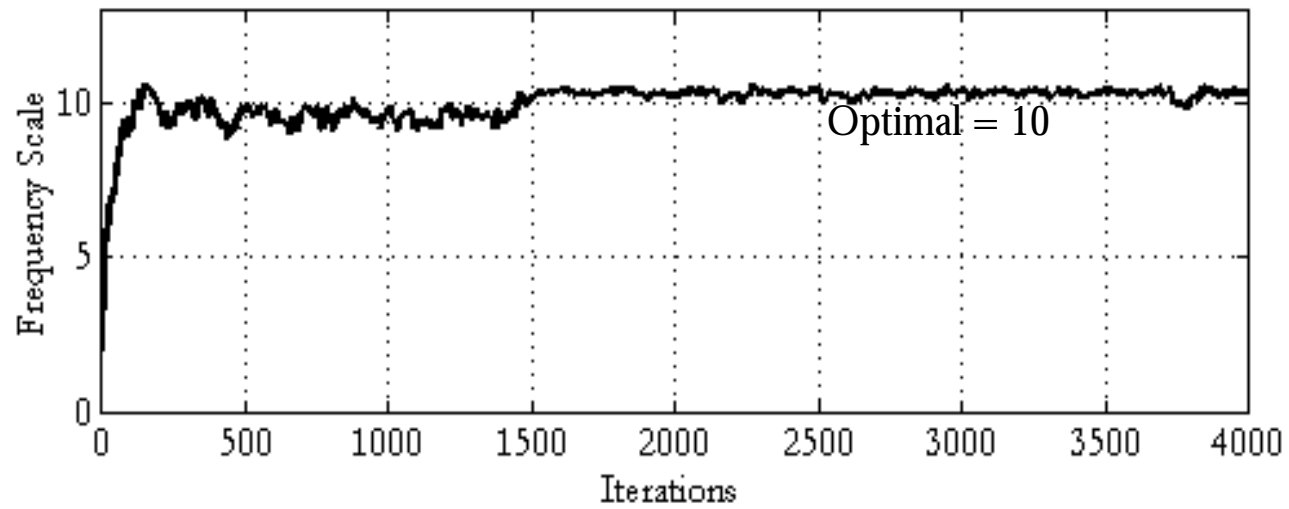
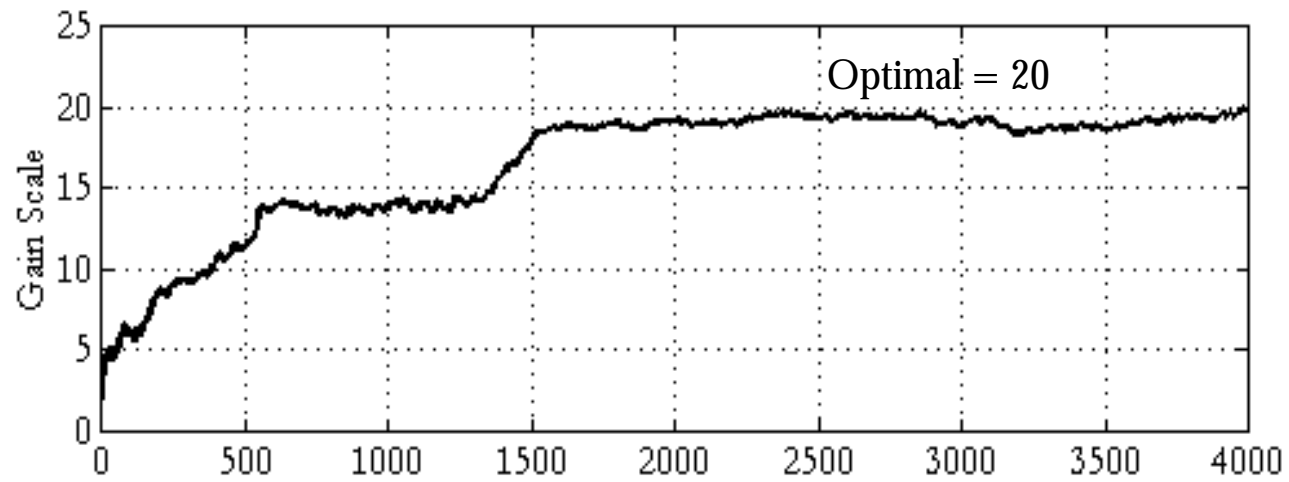


“A 5th order Gm-C filter in 0.25 μm CMOS with digitally programmable poles & zeros,” Chan Carusone & Johns, ISCAS '02, Ballroom G

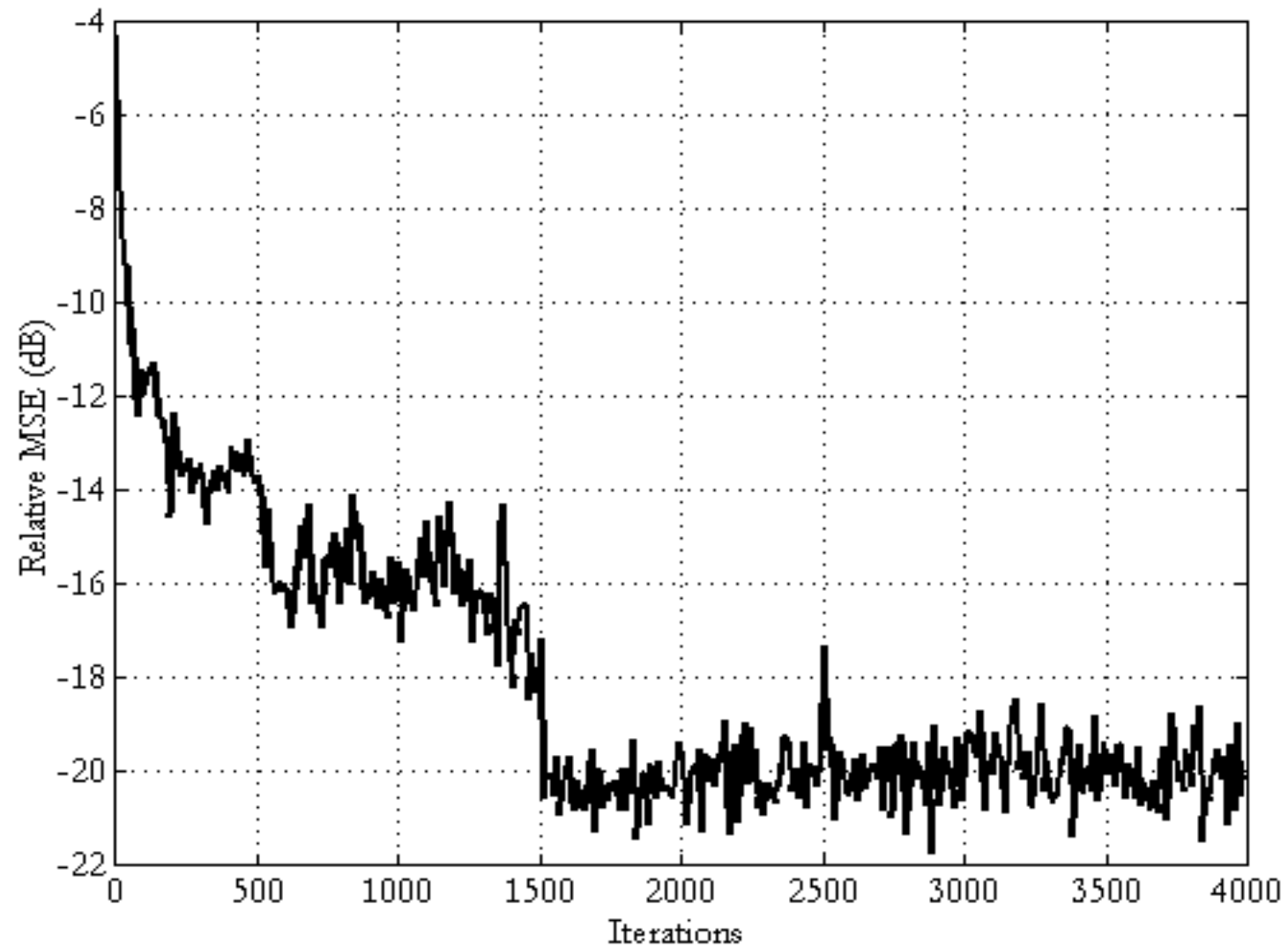
Experimental Setup



Results



Results



Conclusions

- The DLS algorithm is suitable for integrated analog adaptive filters, especially under the following conditions:
 1. the gradient signals required for LMS adaptation are difficult to obtain
 2. slow convergence can be tolerated
- analyses and experimental verification of the algorithm was performed on a 5th order continuous-time integrated filter