

# UNIFIED PRINCIPLE OF $\Sigma\Delta$ MODULATION AS QUANTIZATION TECHNIQUE FOR OVERCOMPLETE EXPANSIONS

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# DISCRETIZATION OF SIGNAL

$\{\Phi_n\}_{n \in \mathbb{Z}}$  : generating family of vectors spanning input space

$$\mathbf{x} = \sum_{n \in \mathbb{Z}} x_n \cdot \Phi_n$$

$$\hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} q_n \cdot \Phi_n \quad \text{where } q_n \in \underbrace{\{l_1, l_2, \dots, l_N\}}_{\text{quantization levels}}$$

**error**  $\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} (x_n - q_n) \cdot \Phi_n$

# USE OF REDUNDANCY IN $\Sigma\Delta$ MODULATION

Form vectors

$$\mathbf{r}_k := \Phi_k - \sum_{n \neq k} c_{n,k} \Phi_n, \quad k \in \mathbb{Z}$$

General notation

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n,k} \Phi_n, \quad k \in \mathbb{Z}, \quad \text{with } d_{k,k} = 1$$

$$D := \{d_{n,k}\}_{n,k \in \mathbb{Z}} : \text{redundancy operator}$$

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} (x_n - q_n) \cdot \Phi_n \quad \Leftrightarrow \quad \mathbf{x} - \hat{\mathbf{x}} = \sum_{k \in \mathbb{Z}} u_k \cdot \mathbf{r}_k$$

where  $x_n - q_n = \sum_{k \in \mathbb{Z}} d_{n,k} \cdot u_k$

# PROOF

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in Z} (x_n - q_n) \cdot \Phi_n$$

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$$x_n - q_n = \sum_{k \in Z} d_{n,k} \cdot u_k \quad \mathbf{r}_k = \sum_{n \in Z} d_{n,k} \cdot \Phi_n$$

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in Z} \underbrace{\left( \sum_{k \in Z} d_{n,k} \cdot u_k \right)}_{x_n - q_n} \cdot \Phi_n = \sum_{k \in Z} u_k \cdot \underbrace{\left( \sum_{n \in Z} d_{n,k} \cdot \Phi_n \right)}_{\mathbf{r}_k}$$

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$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{k \in Z} u_k \cdot \mathbf{r}_k$$

# DIFFERENTIATION EXAMPLE

Form vectors

$$\mathbf{r}_k := \Phi_k - \Phi_{k+1}$$

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in Z} (x_n - q_n) \cdot \Phi_n \quad \Leftrightarrow \quad \mathbf{x} - \hat{\mathbf{x}} = \sum_{k \in Z} u_k \cdot \mathbf{r}_k$$

where  $x_n - q_n = u_n - u_{n-1}$

# DIFFERENTIATION EXAMPLE

Form vectors

$$\mathbf{r}_k := \Phi_k - \Phi_{k+1}$$

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n,k} \Phi_n, \quad k \in \mathbb{Z},$$

where  $d_{n,k} = d_{n-k}$  with  $d_n := \delta_n - \delta_{n+1}$

$$D := \{d_{n-k}\}_{n,k \in \mathbb{Z}}$$

particular case of “convolutional” or “shift-invariant” operator

# PRINCIPLES OF $\Sigma\Delta$ MODULATION

Choose redundancy operator  $D = \{d_{n,k}\}_{n,k \in \mathbb{Z}}$  invertible and

such that  $\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n,k} \Phi_n$  are “small”

Find quantized sequence  $q_n \in \{l_1, l_2, \dots, l_N\}$  so that equation

$$x_n - q_n = \sum_{k \in \mathbb{Z}} d_{n,k} \cdot u_k$$

yields bounded and “small” solution in  $u_k$

Hopefully,  $\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} u_n \cdot \mathbf{r}_n$  will be “small”

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# SPACE NORM

Main case of study: shift-invariant space of functions

$$\Phi_n = \varphi(t - n\tau) \text{ where } \varphi(t) \text{ and } \tau \text{ are given}$$

- $\left\| \mathbf{x} \right\|_2^2 = \int_{-\infty}^{\infty} |x(t)|^2 dt$  however  $\hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} q_n \cdot \Phi_n \notin L^2$   
use of L2 norm requires special treatment, see [O.Yilmaz, 2004]
- $\left\| \mathbf{x} \right\|_{\infty} = \sup_{t \in \mathbf{R}} |x(t)|$
- $\text{MSE}(\mathbf{x}) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T}^{T} |x(t)|^2 dt$

# SPACE NORM

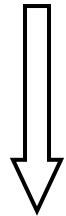
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# SUP NORM BOUND

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in Z} u_n \cdot \mathbf{r}_n$$



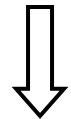
$$|x(t) - \hat{x}(t)| \leq \|u\|_\infty \cdot \sum_{k \in Z} |r_k(t)|$$

[ I.Daubechies & R.DeVore, 2003 ]

# DIFFERENTIATION

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad d_n = \delta_n - \delta_{n-1}$$

$$\mathbf{r}_k := \Phi_k - \Phi_{k+1} \quad \text{with} \quad \Phi_n = \varphi(t - n\tau)$$



$$\sum_{k \in \mathbb{Z}} |r_k(t)| \leq \|\varphi'\|_{L^1}$$

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$$\|\mathbf{x} - \hat{\mathbf{x}}\|_\infty \leq \|u\|_\infty \cdot \|\varphi'\|_{L^1}$$

[ I.Daubechies & R.DeVore, 2003 ]

# APPLICATION TO BANDLIMITED INPUTS

Assumption :  $x(t)$  lowpass of bandwidth  $\Omega_0 = \frac{2\pi}{\tau_0}$

Then  $x(t) := \sum_{n \in \mathbb{Z}} x_n \cdot \varphi(t - n\tau)$

where  $x_n = x(n\tau)$  and  $\varphi(t) = \tau \operatorname{sinc}_{\tau_0}(t)$

with  $\operatorname{sinc}_{\tau_0}(t) := \frac{\sin(\pi t / \tau_0)}{\pi t}$  and  $\tau < \tau_0 := \frac{2\pi}{\Omega_0} \rightarrow R = \frac{\tau_0}{\tau} = \text{redundancy}$

↑  
up to some relaxation in  
the frequency domain, see \*

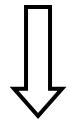
$$\|\mathbf{x} - \hat{\mathbf{x}}\|_\infty \leq \tau \cdot \|\mathbf{u}\|_\infty \cdot \|\operatorname{sinc}_{\tau_0}'\|_{L^1}$$

[\* I.Daubechies & R.DeVore, 2003 ]

# $m^{\text{th}}$ ORDER DIFFERENTIATION

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad d_n = (\delta_n - \delta_{n-1})^{(m)}$$

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n-k} \Phi_n \quad \text{with} \quad \Phi_n = \varphi(t - n\tau)$$



$$\sum_{k \in \mathbb{Z}} |r_k(t)| \leq \tau^{m-1} \|\varphi^{(m)}\|_{L^1}$$

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$$|x(t) - \hat{x}(t)| \leq \|u\|_\infty \cdot \sum_{k \in \mathbb{Z}} |r_k(t)|$$

[ I.Daubechies & R.DeVore, 2003 ]

# APPLICATION TO BANDLIMITED INPUTS

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad d_n = (\delta_n - \delta_{n-1})^{(m)}$$

$$\sum_{k \in \mathbb{Z}} |r_k(t)| \leq \tau^{m-1} \left\| \text{sinc}_{\tau_0}^{(m)} \right\|_{L^1}$$

---

$$\|\mathbf{x} - \hat{\mathbf{x}}\|_\infty \leq \tau^m \cdot \|u\|_\infty \cdot \left\| \text{sinc}_{\tau_0}^{(m)} \right\|_{L^1}$$

[ I.Daubechies & R.DeVore, 2003 ]

# SPACE NORM

Main case of study: shift-invariant space of functions

$$\Phi_n = \varphi(t - n\tau) \text{ where } \varphi(t) \text{ and } \tau \text{ are given}$$

- 
- $\|\mathbf{x}\|_\infty = \sup_{t \in \mathbf{R}} |x(t)|$
- $\text{MSE}(\mathbf{x}) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T}^T |x(t)|^2 dt$

# MSE ANALYSIS

Main case of study: shift-invariant space of functions  
shift-invariant operator  $D$

$\Phi_n = \varphi(t - n\tau)$  where  $\varphi(t)$  and  $\tau$  are given

$\Rightarrow \mathbf{r}_k = r(t - k\tau)$  where  $r(t) = \sum_{n \in \mathbb{Z}} d_n \varphi(t - n\tau)$

Theorem:

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{k \in \mathbb{Z}} u_k \cdot \mathbf{r}_k \quad \Rightarrow$$

$$\text{MSE}(\mathbf{x} - \hat{\mathbf{x}}) = \sum_{k \in \mathbb{Z}} a_k s_k$$

where

$$a_k := \lim_{N \rightarrow \infty} \frac{1}{2N} \sum_{n=-N}^N u_n u_{n+k} \quad \text{and} \quad s_k := \frac{1}{\tau} \int_{\mathbf{R}} r(t) r(t - k\tau) dt$$

# STATISTICAL MSE ANALYSIS

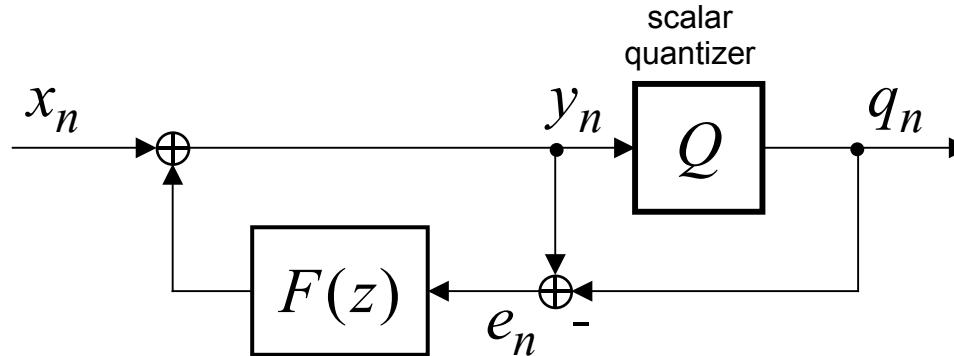
$$E\{\text{MSE}(\mathbf{x} - \hat{\mathbf{x}})\} = \sum_{k \in \mathbb{Z}} \bar{a}_k s_k$$

where

$$\bar{a}_k := E\{u_n u_{n+k}\}$$

$$\text{and } s_k := \frac{1}{\tau} \int_{\mathbf{R}} r(t) r(t - k\tau) dt$$

# “WHITE NOISE” BEHAVIOR



$$u_n = g_n * e_n$$

$$E\{e_n e_{n+k}\} \xrightarrow[\text{large \#bits}]{} \sigma_e^2 \delta_k$$

$$\Rightarrow \bar{a}_k := E\{u_n u_{n+k}\} \xrightarrow[\text{large \#bits}]{} \bar{a}_0 \delta_k$$

# “WHITE NOISE” MODEL

$$E\{\text{MSE}(\mathbf{x} - \hat{\mathbf{x}})\} = \frac{\bar{a}_0}{\tau} \|\mathbf{r}\|_2^2$$

$$\bar{a}_k := E\{u_n u_{n+k}\} = \bar{a}_0 \delta_k$$

$$s_0 = \frac{1}{\tau} \int_{\mathbf{R}} |r(t)|^2 dt = \frac{1}{\tau} \|\mathbf{r}\|_2^2$$

# FIR DESIGN OF $D$ UNDER “WHITE NOISE” MODEL

Assume constraint

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad \{d_n\}_{n \in \mathbb{Z}} = \underbrace{\{d_0, d_1, \dots, d_m\}}_{=1}$$

$$r(t) = \sum_{n \in \mathbb{Z}} d_n \varphi(t - n\tau) \quad \Rightarrow \quad \mathbf{r} = \Phi_0 + d_1 \Phi_1 + \dots + d_m \Phi_m$$

$$E\{\text{MSE}(\mathbf{x} - \hat{\mathbf{x}})\} = \frac{\bar{a}_0}{\tau} \|\mathbf{r}\|_2^2$$

is iminimized when  $d_1, \dots, d_m$  are chosen such that

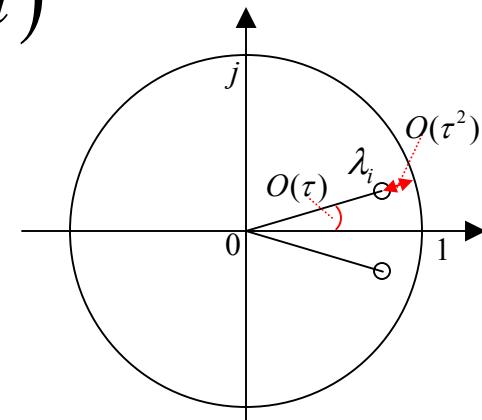
$$d_1 \Phi_1 + \dots + d_m \Phi_m = -\text{proj}_{\langle \Phi_1, \dots, \Phi_m \rangle}(\Phi_0)$$

# BANDLIMITED CASE

$$\varphi(t) = \tau \operatorname{sinc}_{\tau_0}(t)$$

White noise model optimization

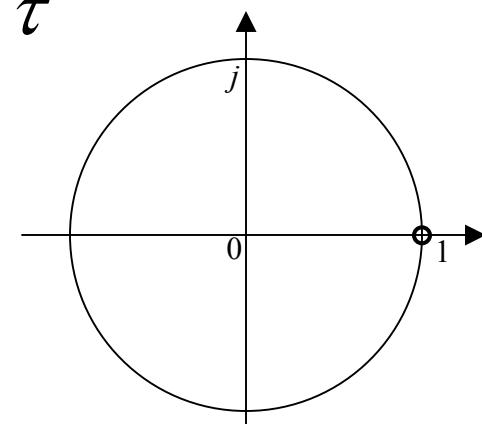
$$\Rightarrow D(z) = \prod_{i=1}^m (1 - \lambda_i z^{-1})$$



$$E\{\text{MSE}(\mathbf{x} - \hat{\mathbf{x}})\} \approx \alpha \tau^{2m+1}$$

Differentiation operator

$$D(z) = (1 - z^{-1})^m$$



$$E\{\text{MSE}(\mathbf{x} - \hat{\mathbf{x}})\} \approx \beta \tau^{2m+1}$$

$$\|\mathbf{x} - \hat{\mathbf{x}}\|_\infty^2 \leq \gamma \tau^{2m}$$

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Find quantized sequence  $q_n \in \{l_1, l_2, \dots, l_N\}$  so that equation

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# CAUSAL AND TIME-INVARIANT FRAMEWORK

- $D = \{d_{n,k}\}_{n,k \in \mathbb{Z}}$  such that  $d_{n,k} = d_{n-k}$  with  $d_n = 0, \forall n < 0$  and  $d_0 = 1$
- $q_n$  must be decided at instant  $n$

Find quantized sequence  $q_n \in \{l_1, l_2, \dots, l_N\}$  so that equation

$$x_n - q_n = d_n * u_n$$

yields bounded and “small” solution in  $u_k$

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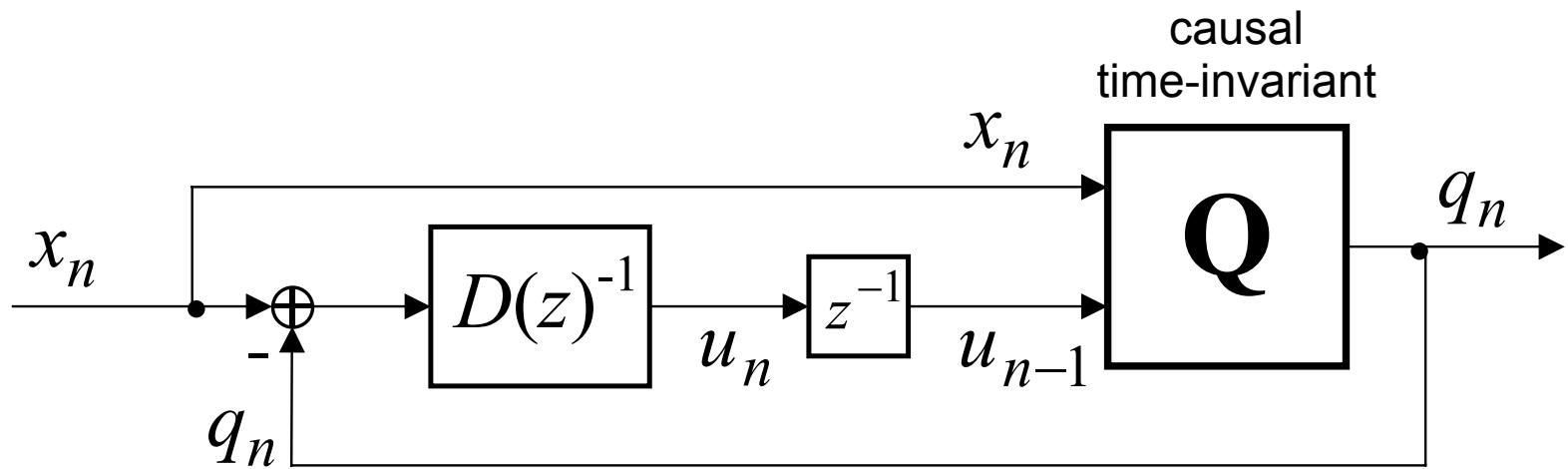
$$u_n = - \sum_{k>0} d_k \cdot u_{n-k} + (x_n - q_n)$$

yields bounded and “small” solution in  $u_k$

$$\begin{aligned} q_n &= \mathbf{Q}(x_n, x_{n-1}, \dots; u_{n-1}, u_{n-2}, \dots) \\ &\in \{l_1, l_2, \dots, l_N\} \end{aligned}$$

# CAUSAL AND TIME-INVARIANT RESOLUTION

dynamical system

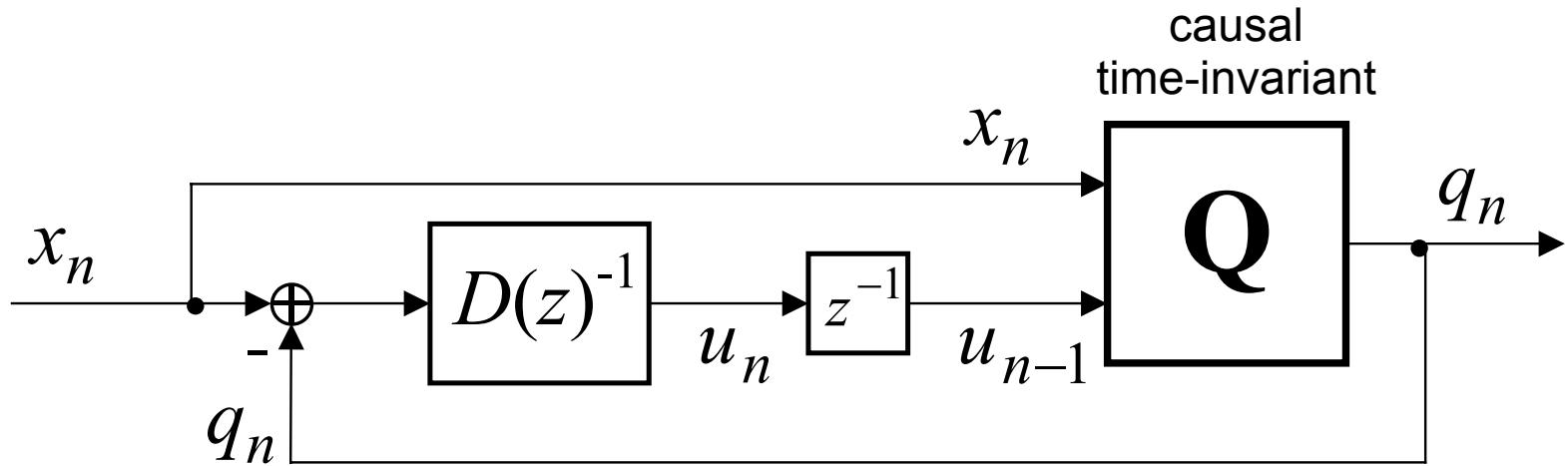


$$u_n = - \sum_{k>0} d_k \cdot u_{n-k} + (x_n - q_n)$$

$$q_n = \mathbf{Q}(x_n, x_{n-1}, \dots; u_{n-1}, u_{n-2}, \dots)$$

# STABLE ONE-BIT SCHEME

[ Daubechies & DeVore (2003) ]

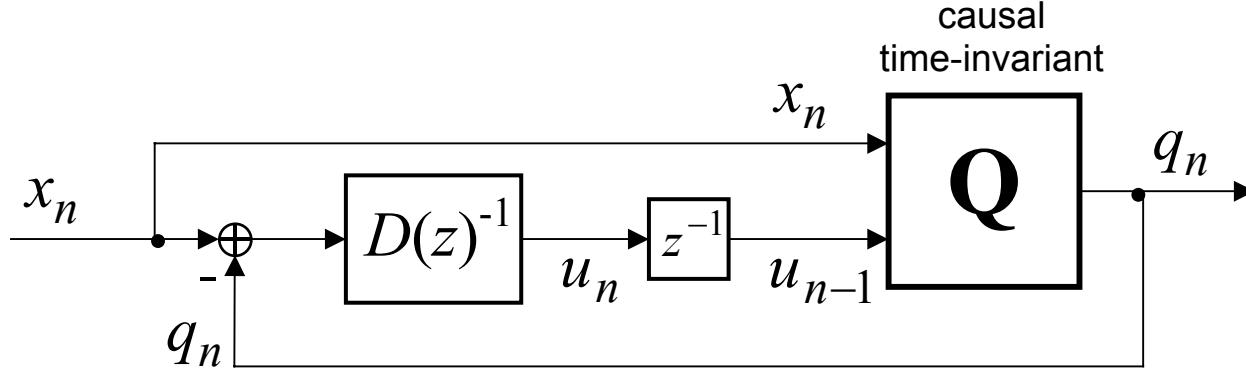


$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \text{ such that } d_n = (\delta_n - \delta_{n-1})^{(m)} \quad (D(z) = (1 - z^{-1})^m)$$

$$|x_n| \leq \frac{\Delta}{2} - \varepsilon$$

$$q_n = \mathbf{Q}(x_n; u_{n-1}, u_{n-2}, \dots, u_{n-m}) \in \left\{ +\frac{\Delta}{2}, -\frac{\Delta}{2} \right\}$$

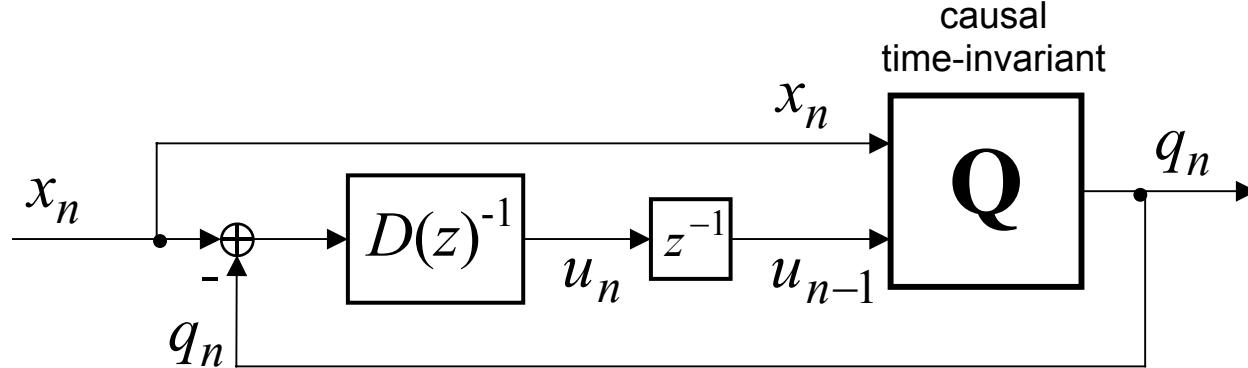
# RESTRICTED SCHEME



restriction:  $Q(x_n, x_{n-1}, \dots; u_{n-1}, u_{n-2}, \dots) = Q(x_n + c_n * u_n)$

where  $c_n$  is strictly causal and  $Q$  is a **scalar** quantizer

# RESTRICTED SCHEME



**restriction:**  $Q(x_n, x_{n-1}, \dots; u_{n-1}, u_{n-2}, \dots) = Q(\underbrace{x_n + c_n * u_n}_{y_n})$

$$q_n = Q(y_n)$$

$$x_n - q_n = d_n * u_n$$

$$y_n = x_n + c_n * u_n$$

# RESTRICTED SCHEME

Define

$$e_n := y_n - q_n$$

Perform change of state variable

$$e_n = (d_n + c_n) * u_n$$

$$q_n = Q(y_n)$$

$$y_n - q_n = (d_n + c_n) * u_n$$

$$y_n = x_n + c_n * u_n$$

# RESTRICTED SCHEME

Define

$$e_n := y_n - q_n$$

Perform change of state variable

$$e_n = (d_n + c_n) * u_n$$

$$u_n = g_n * e_n$$

where  $g_n$  such that  $G(z) = \frac{1}{D(z) + C(z)}$

$$q_n = Q(y_n)$$

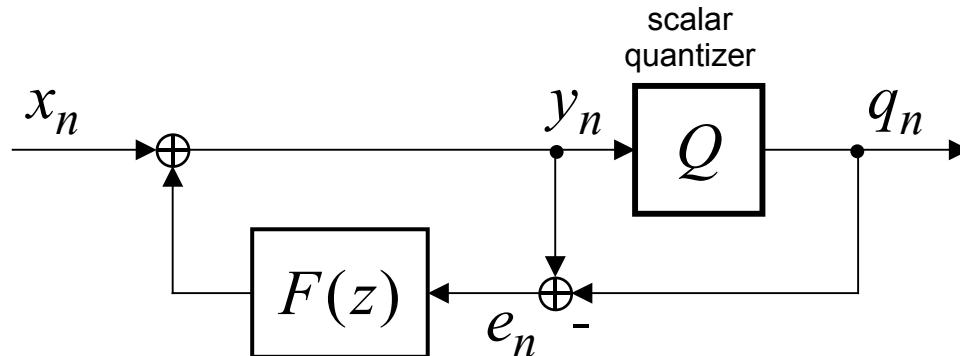
$$q_n = Q(y_n)$$

$$y_n - q_n = (d_n + c_n) * u_n \quad \Leftrightarrow \quad e_n = y_n - q_n$$

$$y_n = x_n + c_n * u_n$$

$$y_n = x_n + \underbrace{c_n * g_n * e_n}_{f_n}$$

# ERROR DIFFUSION



$$u_n = g_n * e_n$$

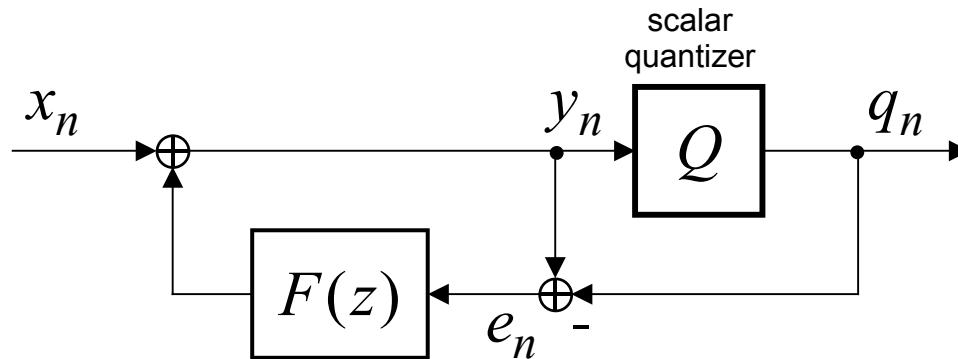
where  $G(z) = \frac{1}{D(z) + C(z)}$  and  $F(z) = C(z)G(z)$

$$q_n = Q(y_n)$$

$$e_n = y_n - q_n$$

$$y_n = x_n + \underbrace{c_n * g_n}_f * e_n$$

# STABLE ERROR DIFFUSION



$$u_n = g_n * e_n$$

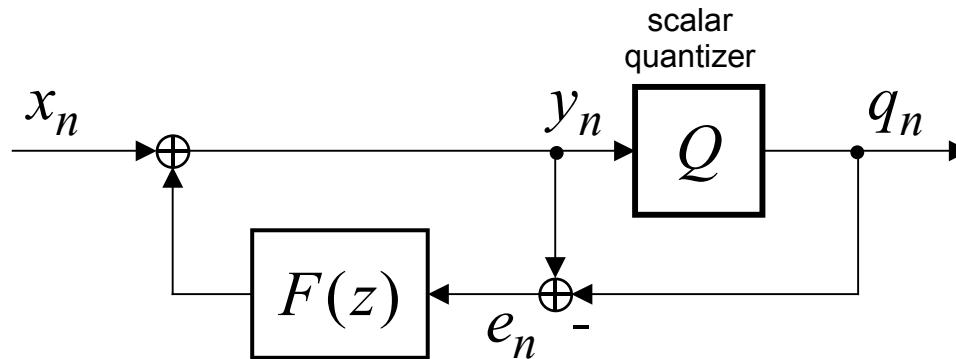
where  $G(z) = \frac{1}{D(z) + C(z)}$  and  $F(z) = C(z)G(z)$

$$q_n = Q(y_n)$$

$$e_n = y_n - q_n$$

$$\|y\|_\infty \leq \|x\|_\infty + \|f\|_1 \cdot \|e\|_\infty \iff y_n = x_n + f_n * e_n$$

# STABLE ERROR DIFFUSION



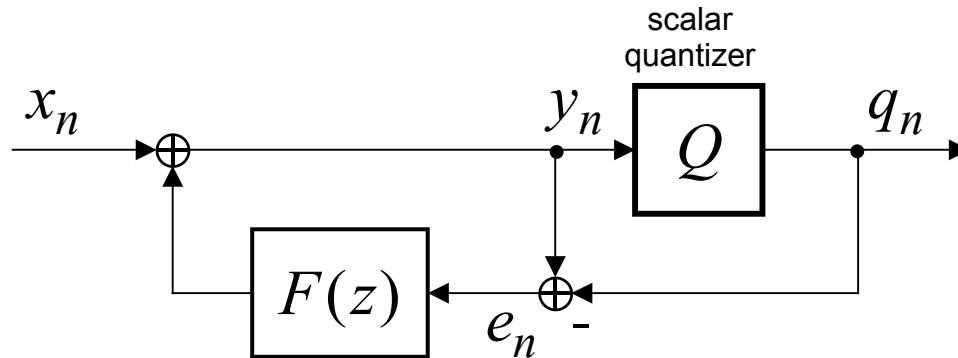
$$u_n = g_n * e_n$$

where  $G(z) = \frac{1}{D(z) + C(z)}$  and  $F(z) = C(z)G(z)$

- choose  $C(z)$  so that both  $G(z)$  and  $F(z)$  are stable
- choose uniform quantizer  $Q$  of step size  $\Delta$   $\Rightarrow \|e\|_\infty \leq \frac{\Delta}{2}$

$$\|y\|_\infty \leq \|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \quad \Rightarrow \quad \text{finite quantizer } Q$$

# STABLE 1-BIT ERROR DIFFUSION

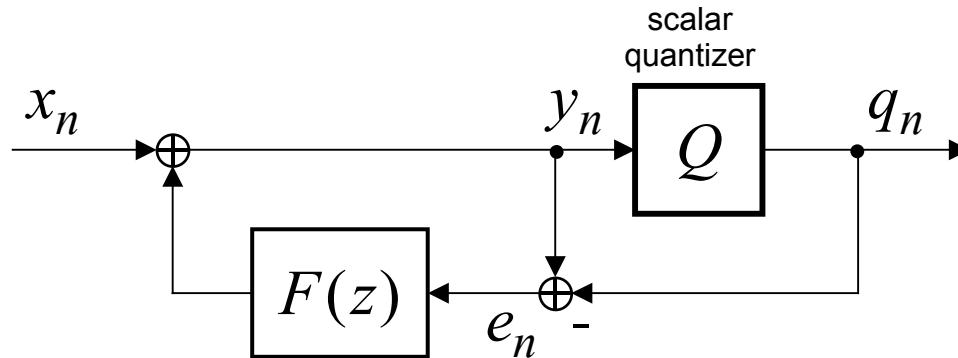


$$u_n = g_n * e_n$$

where  $G(z) = \frac{1}{D(z) + C(z)}$  and  $F(z) = C(z)G(z)$

- choose  $C(z)$  so that both  $G(z)$  and  $F(z)$  are stable
- choose uniform quantizer  $Q$  of step size  $\Delta \Rightarrow \|e\|_\infty \leq \frac{\Delta}{2}$
- design  $f_n$  so that  $\|y\|_\infty \leq \|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta \Rightarrow$  1-bit quantizer  $Q$

# STABLE 1-BIT ERROR DIFFUSION



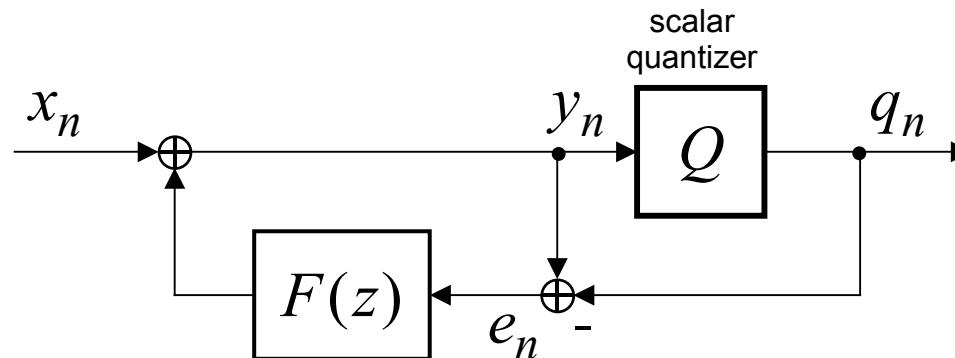
$$u_n = g_n * e_n$$

where  $G(z) = \frac{1}{D(z) + C(z)}$  and  $F(z) = C(z)G(z)$

$$\Rightarrow g_n \text{ causal with } g_0=1 \text{ and } F(z) = 1 - G(z)D(z)$$

- design  $g_n$  so that  $\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta$   $\Rightarrow$  1-bit quantizer  $Q$

# CASE OF DIFFERENTIATOR



$$D(z) = (1 - z^{-1})^m$$

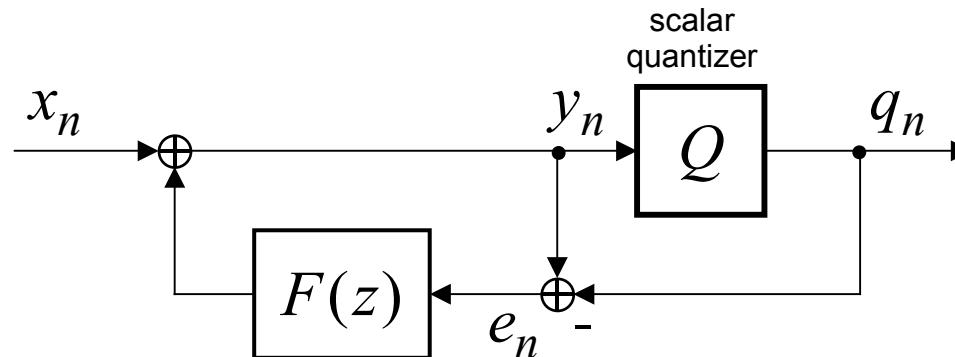
$$F(z) = 1 - G(z)D(z)$$

choose  $G(z) = 1$       then       $F(z) = 1 - D(z)$

$$\|f_1\| = 2^m - 1$$

$$\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta \iff \begin{cases} m = 1 \\ \|x\|_\infty \leq \frac{\Delta}{2} \end{cases}$$

# SINGLE-LOOP SINGLE-BIT CASE



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

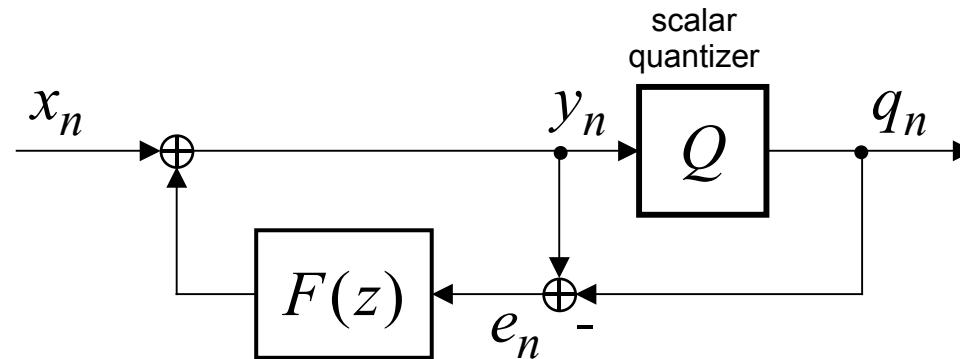
$$F(z) = 1 - G(z)D(z)$$

choose  $G(z) = 1$       then       $F(z) = z^{-1}$

$$\|f_1\| = 1$$

$$\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta \iff \begin{cases} m = 1 \\ \|x\|_\infty \leq \frac{\Delta}{2} \end{cases}$$

# CASE OF DIFFERENTIATOR WITH $m > 1$

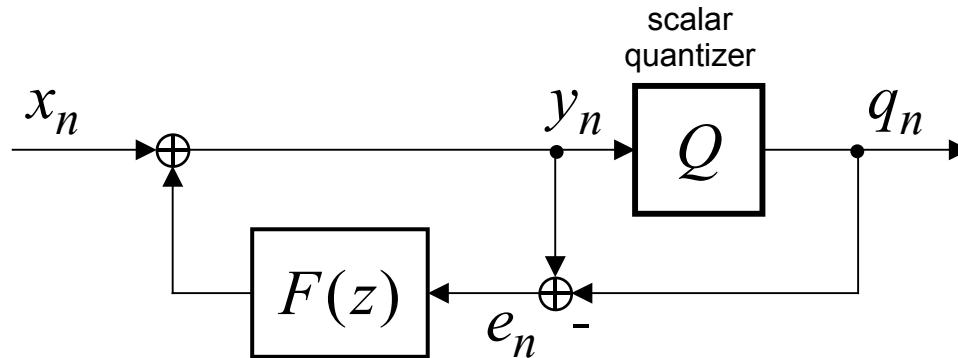


$$D(z) = (1 - z^{-1})^m$$

$$F(z) = 1 - G(z)D(z)$$

$$\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta \quad \Rightarrow \quad G(z) \neq 1$$

# DIFFERENTIATOR WITH FIR $G(z)$



$$D(z) = (1 - z^{-1})^m$$

$$F(z) = 1 - G(z)D(z)$$

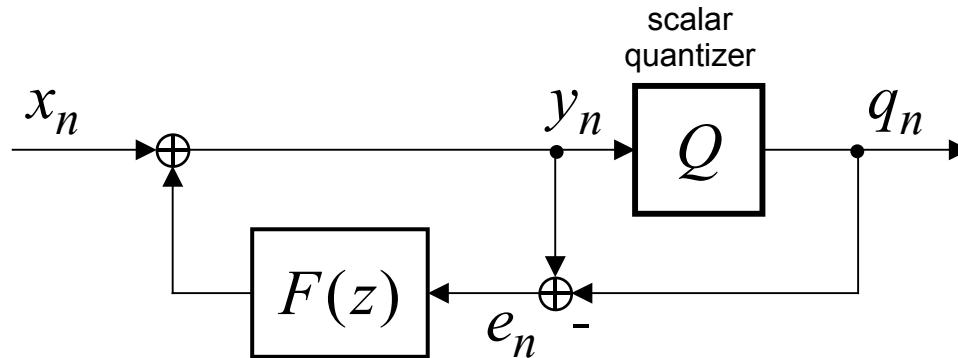
Assumption:  $\|x\|_\infty \leq \frac{\Delta}{2} - \varepsilon$

There exists FIR filter  $G(z)$  such that  $\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta$

$$\text{length}(g_n) \approx 6m^2$$

[ S.Gunturk, 2003 ]

# DIFFERENTIATOR WITH IIR $G(z)$



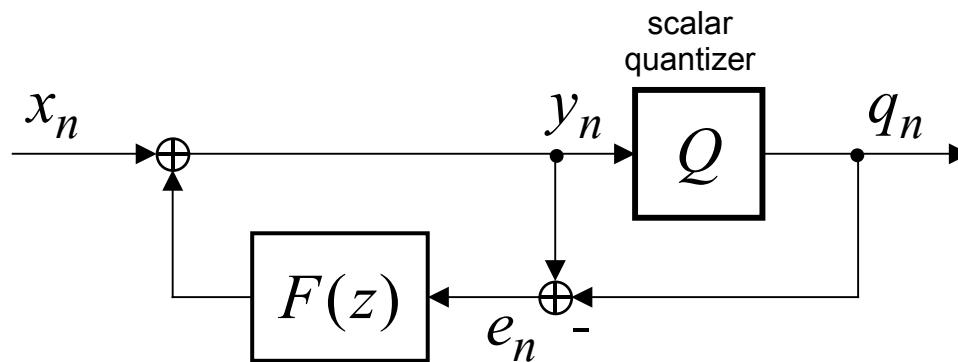
$$D(z) = (1 - z^{-1})^m$$

$$F(z) = 1 - G(z)D(z)$$

with  $G(z)$  of the type  $G(z) = \frac{1}{1 + a_1 z^{-1} + \dots + a_m z^{-m}}$

- no existing analytical method to guarantee  $\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta$
- in practice, however, most popular 1-bit scheme !
- empirical design method established by R. Schreier (1993) that leads to (overloaded) stability and high performances

# DIFFERENTIATOR WITH IIR $G(z)$ : 2nd ORDER



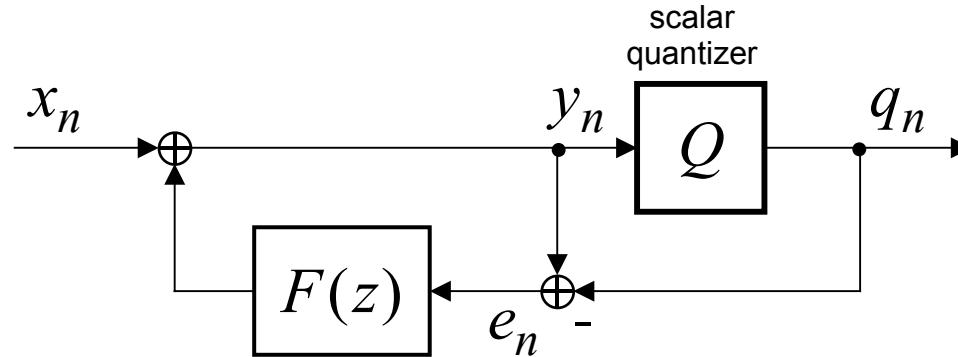
$$D(z) = (1 - z^{-1})^2$$

$$F(z) = 1 - G(z)D(z)$$

with  $G(z)$  of the type  $G(z) = \frac{1}{1 + a_1 z^{-1} + a_2 z^{-2}}$

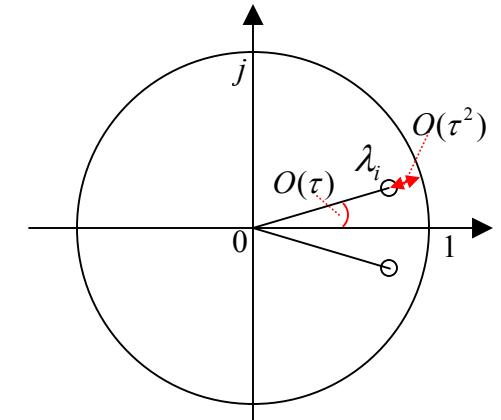
Rigorous proof of (overloaded) stability by [O.Yilmaz, 2002]

# LOWPASS $\Sigma\Delta$ WITH IIR $G(z)$



$$D(z) = \prod_{i=1}^m (1 - \lambda_i z^{-1})$$

$$F(z) = 1 - G(z) D(z)$$



with  $G(z)$  of the type  $G(z) = \frac{1}{1 + a_1 z^{-1} + \dots + a_m z^{-m}}$

- no existing analytical method to guarantee  $\|x\|_\infty + \|f\|_1 \cdot \frac{\Delta}{2} \leq \Delta$
- in practice, however, most popular 1-bit scheme !
- empirical design method established by R. Schreier (1993) that leads to (overloaded) stability and high performances

# PRINCIPLES OF $\Sigma\Delta$ MODULATION

Choose redundancy operator  $D = \{d_{n,k}\}_{n,k \in \mathbb{Z}}$  invertible and

such that  $\mathbf{r}_k := \sum_{n \neq k} d_{n,k} \Phi_n$  are “small”

Find quantized sequence  $q_n \in \{l_1, l_2, \dots, l_N\}$  so that equation

$$x_n - q_n = \sum_{k \in \mathbb{Z}} d_{n,k} \cdot u_k$$

yields bounded and “small” solution in  $u_k$

Hopefully,  $\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} u_n \cdot \mathbf{r}_n$  will be “small”

# VARIOUS APPLICATIONS

- Bandpass  $\Sigma\Delta$  modulation
- Multi-channel  $\Sigma\Delta$  modulation
- 2D  $\Sigma\Delta$  modulation (image halftoning, time-frequency)
- Finite dimensional space  $\Sigma\Delta$  modulation

# BANDPASS $\Sigma\Delta$ MODULATION

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad d_n = (\delta_n - \delta_{n-1})^{(m)}$$

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n-k} \Phi_n \quad \text{with} \quad \Phi_n = \varphi(t - n\tau)$$

$\Downarrow$

$$\varphi(t) = \cos(\Omega_0 t) \psi(t)$$

↑  
lowpass

$$\sum_{k \in \mathbb{Z}} |r_k(t)| \leq \tau^{m-1} \|\varphi^{(m)}\|_{L^1}$$

---

$$|x(t) - \hat{x}(t)| \leq \|u\|_\infty \cdot \sum_{k \in \mathbb{Z}} |r_k(t)|$$

# BANDPASS $\Sigma\Delta$ MODULATION

$$D = \{d_{n-k}\}_{n,k \in \mathbb{Z}} \quad \text{with} \quad d_n = (\delta_n - e^{j\omega_0} \delta_{n-1})^{(m)} * (\delta_n - e^{-j\omega_0} \delta_{n-1})^{(m)}$$

and  $\omega_0 = \Omega_0 \tau$

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n-k} \Phi_n \quad \text{with} \quad \Phi_n = \varphi(t - n\tau)$$

$\Downarrow$

$$\varphi(t) = \cos(\Omega_0 t) \psi(t)$$

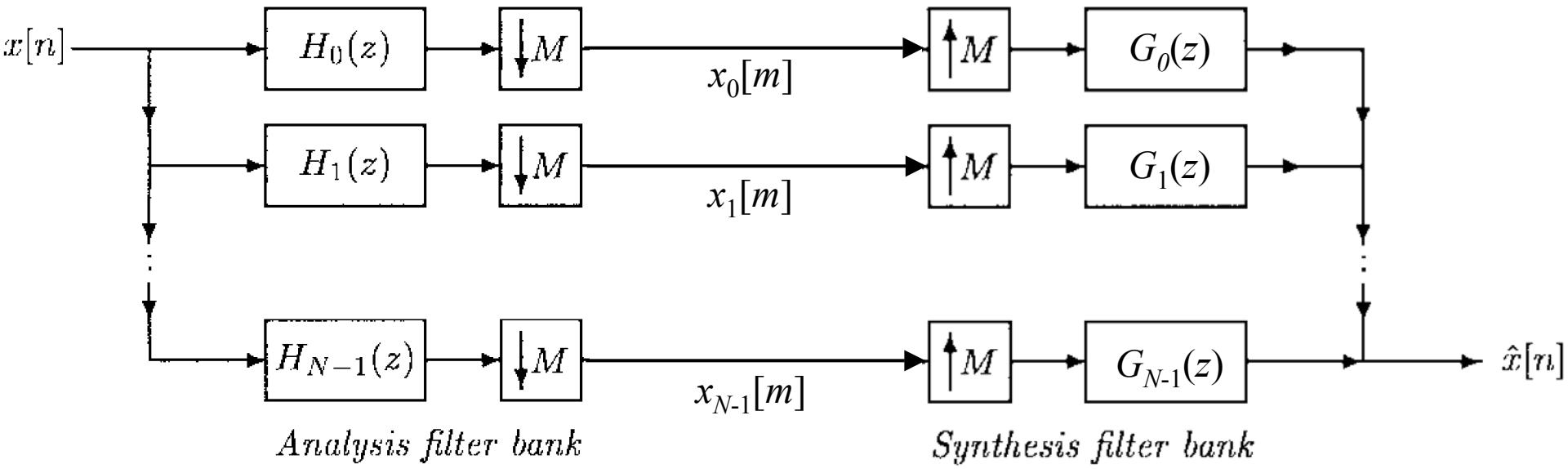
$\uparrow$   
lowpass

$$\sum_{k \in \mathbb{Z}} |r_k(t)| \leq (2\tau)^{m-1} \|\psi^{(m)}\|_{L^1}$$

---

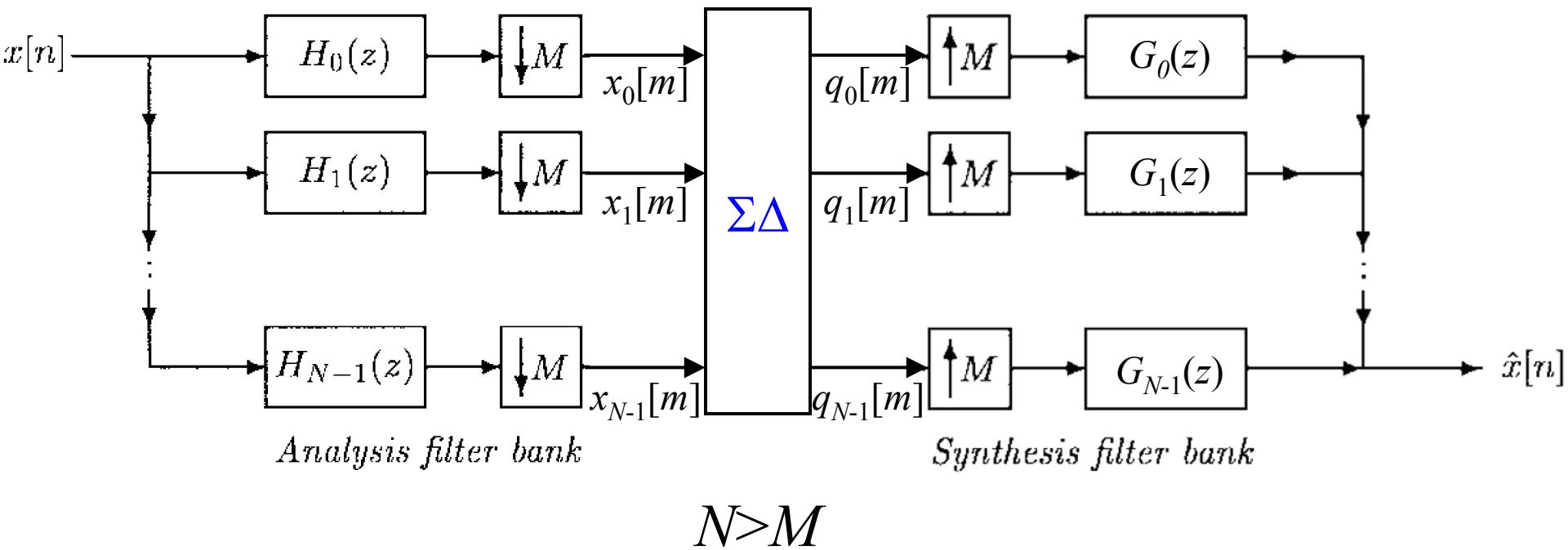
$$|x(t) - \hat{x}(t)| \leq \|u\|_\infty \cdot \sum_{k \in \mathbb{Z}} |r_k(t)|$$

# MULTI-CHANNEL $\Sigma\Delta$ MODULATION

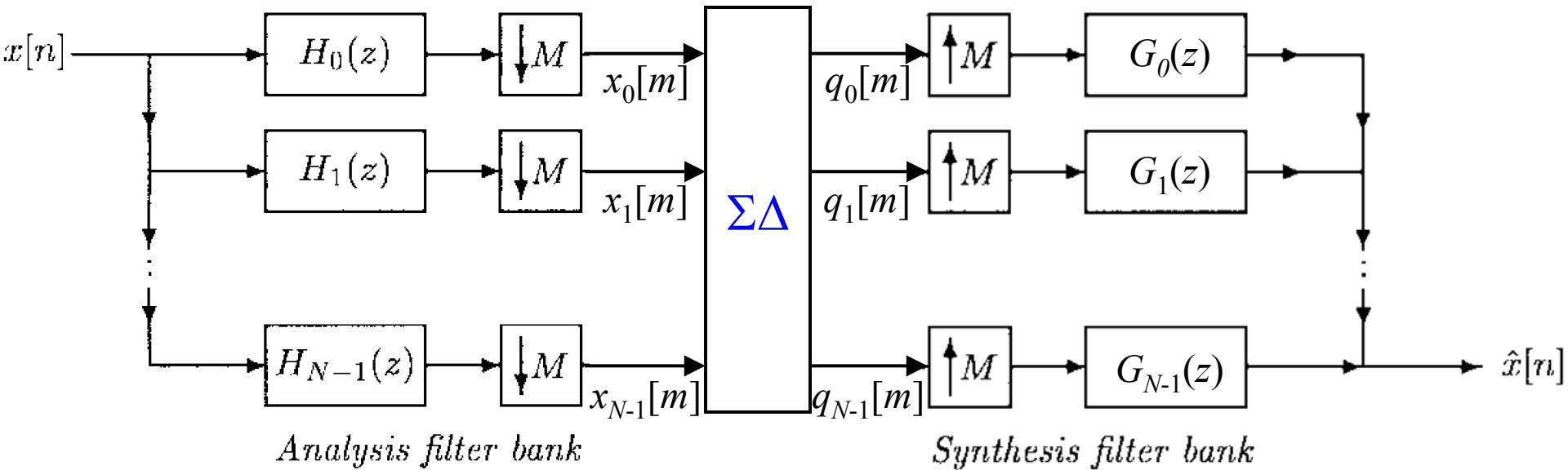


oversampling when  $N > M$

# MULTI-CHANNEL $\Sigma\Delta$ MODULATION



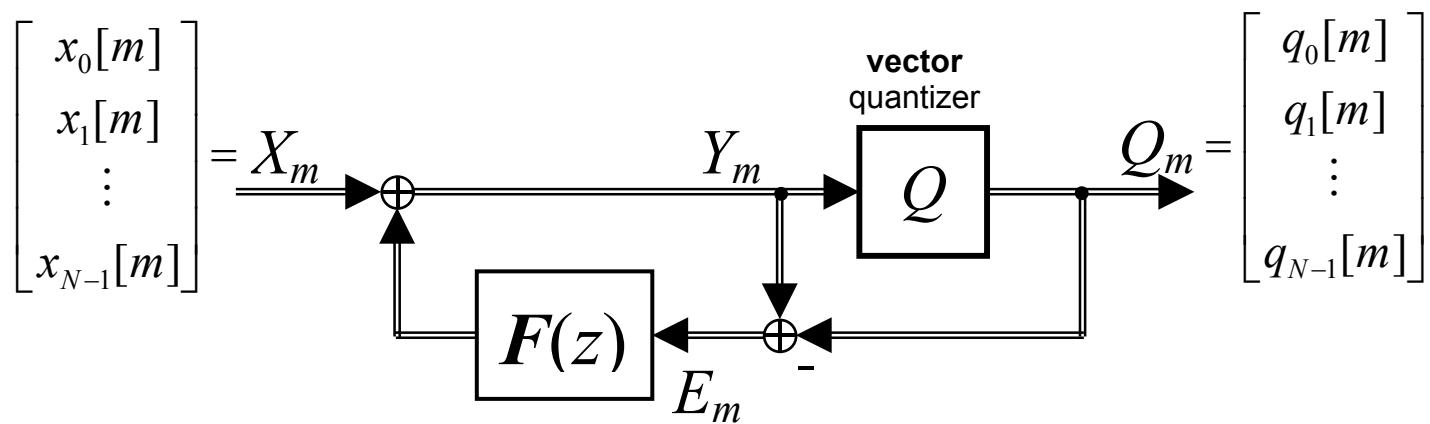
# MULTI-CHANNEL $\Sigma\Delta$ MODULATION



*Analysis filter bank*

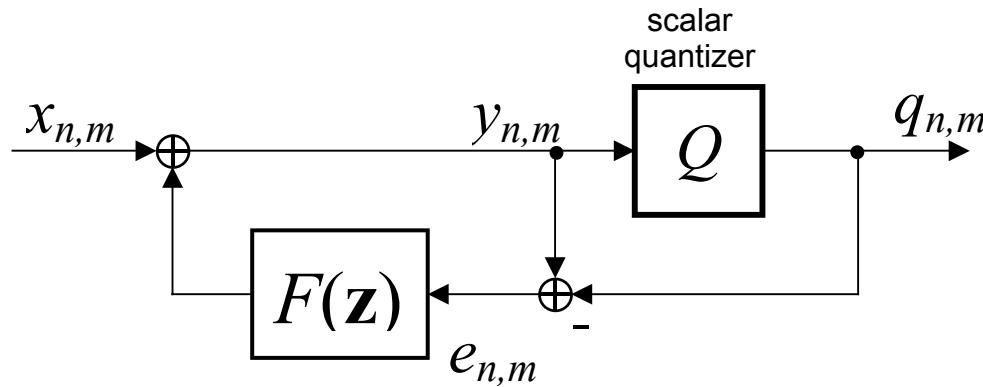
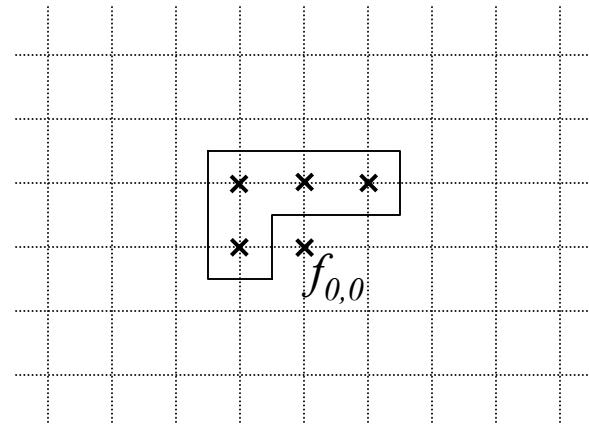
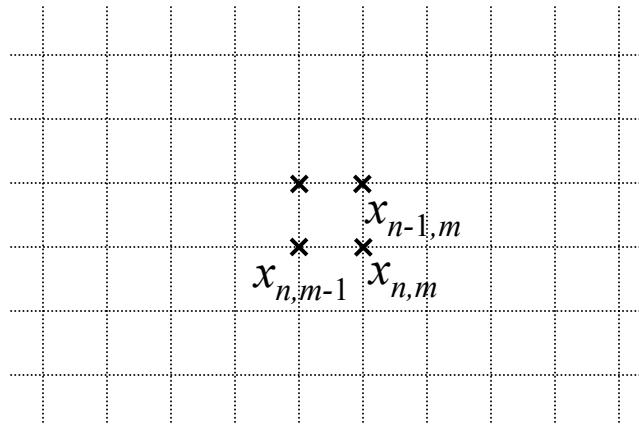
*Synthesis filter bank*

$N > M$



[ H.Bolcskei, 2001 ]

# MULTI-DIMENSIONAL $\Sigma\Delta$ MODULATION



[ Image halftoning: C.W.Wu ]

[ Time-frequency analysis: O.Yilmaz, 2004 ]

# INFINITE DIMENSIONAL SPACE

$$\mathbf{x} = \sum_{n \in \mathbb{Z}} x_n \cdot \Phi_n$$

$$\hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} q_n \cdot \Phi_n$$

$$\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n-k} \cdot \Phi_n$$

$$x_n - q_n = \sum_{k \in \mathbb{Z}} d_{n-k} \cdot u_k$$

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} (x_n - q_n) \cdot \Phi_n \quad \Leftrightarrow \quad \mathbf{x} - \hat{\mathbf{x}} = \sum_{k \in \mathbb{Z}} u_k \cdot \mathbf{r}_k$$

# FINITE DIMENSIONAL SPACE

$$\mathbf{x} = \sum_{n=1}^N x_n \cdot \Phi_n$$

$$\hat{\mathbf{x}} = \sum_{n=1}^N q_n \cdot \Phi_n$$

$$\mathbf{r}_k := \sum_{n=1}^N d_{n-k} \cdot \Phi_n, \quad k = 1, \dots, N$$

$$x_n - q_n = \sum_{k=1}^N d_{n-k} \cdot u_k, \quad n = 1, \dots, N$$

$$\mathbf{x} - \hat{\mathbf{x}} = \sum_{n=1}^N (x_n - q_n) \cdot \Phi_n \iff \mathbf{x} - \hat{\mathbf{x}} = \sum_{n=1}^N u_k \cdot \mathbf{r}_k + B$$

↑  
boundary  
term

[ J.Benedetto, O.Yilmaz & A.Powell, 2005 ]

# PRINCIPLES OF $\Sigma\Delta$ MODULATION

Choose redundancy operator  $D = \{d_{n,k}\}_{n,k \in \mathbb{Z}}$  invertible and

such that  $\mathbf{r}_k := \sum_{n \in \mathbb{Z}} d_{n,k} \Phi_n$  are “small”

Find quantized sequence  $q_n \in \{l_1, l_2, \dots, l_N\}$  so that equation

$$x_n - q_n = \sum_{k \in \mathbb{Z}} d_{n,k} \cdot u_k$$

yields bounded and “small” solution in  $u_k$

Hopefully,  $\mathbf{x} - \hat{\mathbf{x}} = \sum_{n \in \mathbb{Z}} u_n \cdot \mathbf{r}_n$  will be “small”