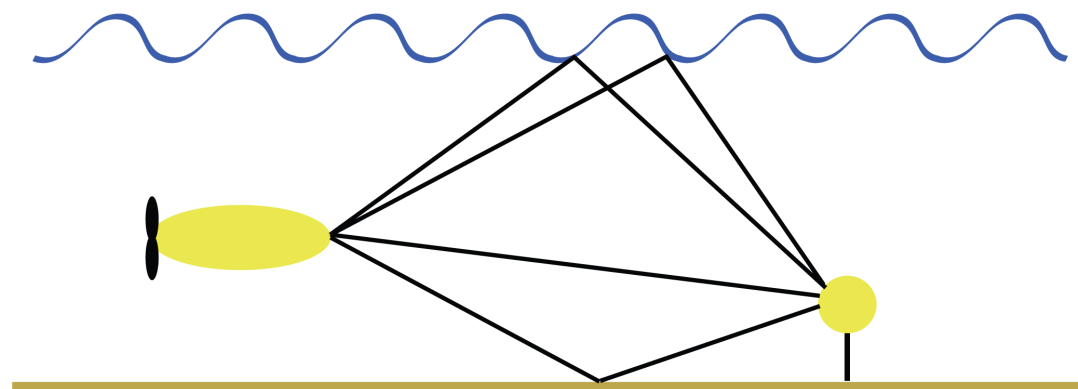


Comparison and Analysis of Equalization Techniques for the Time-Varying Underwater Acoustic Channel



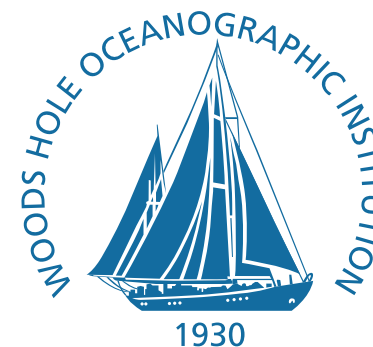
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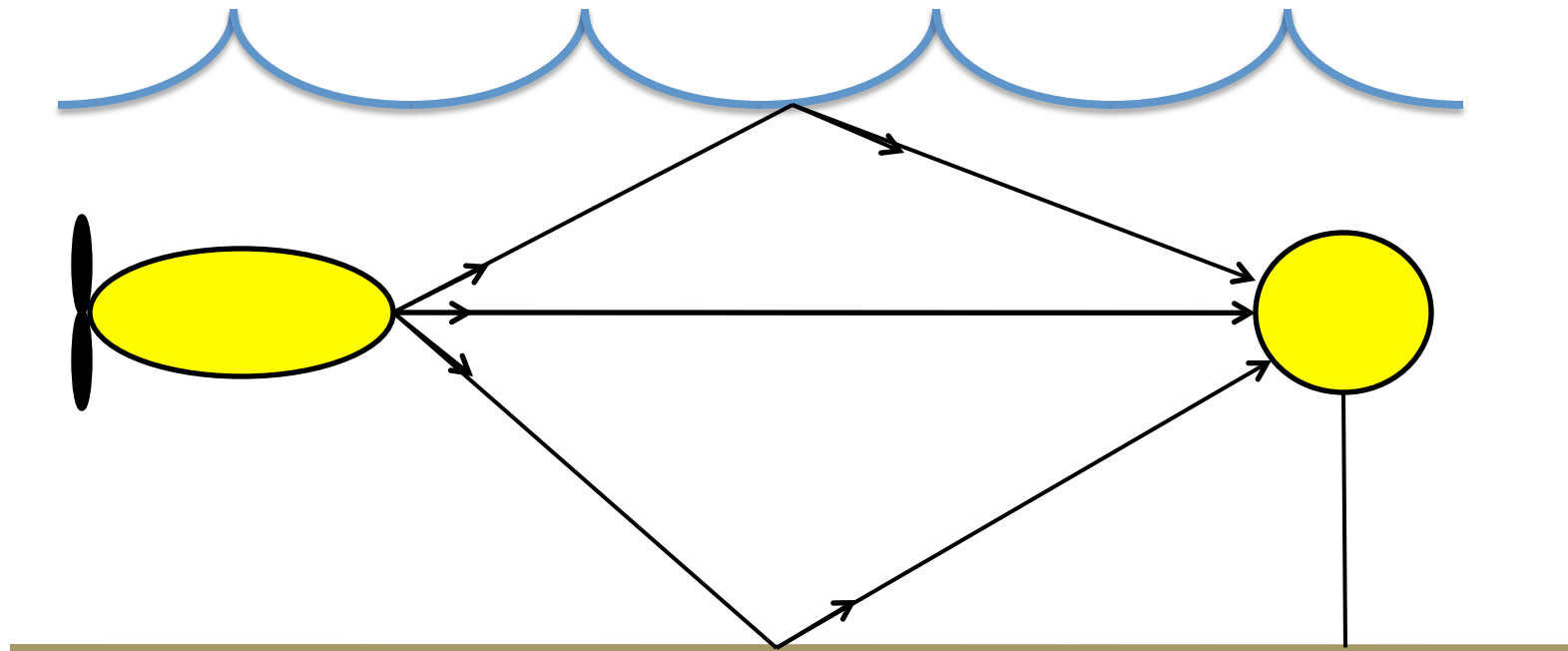




Outline

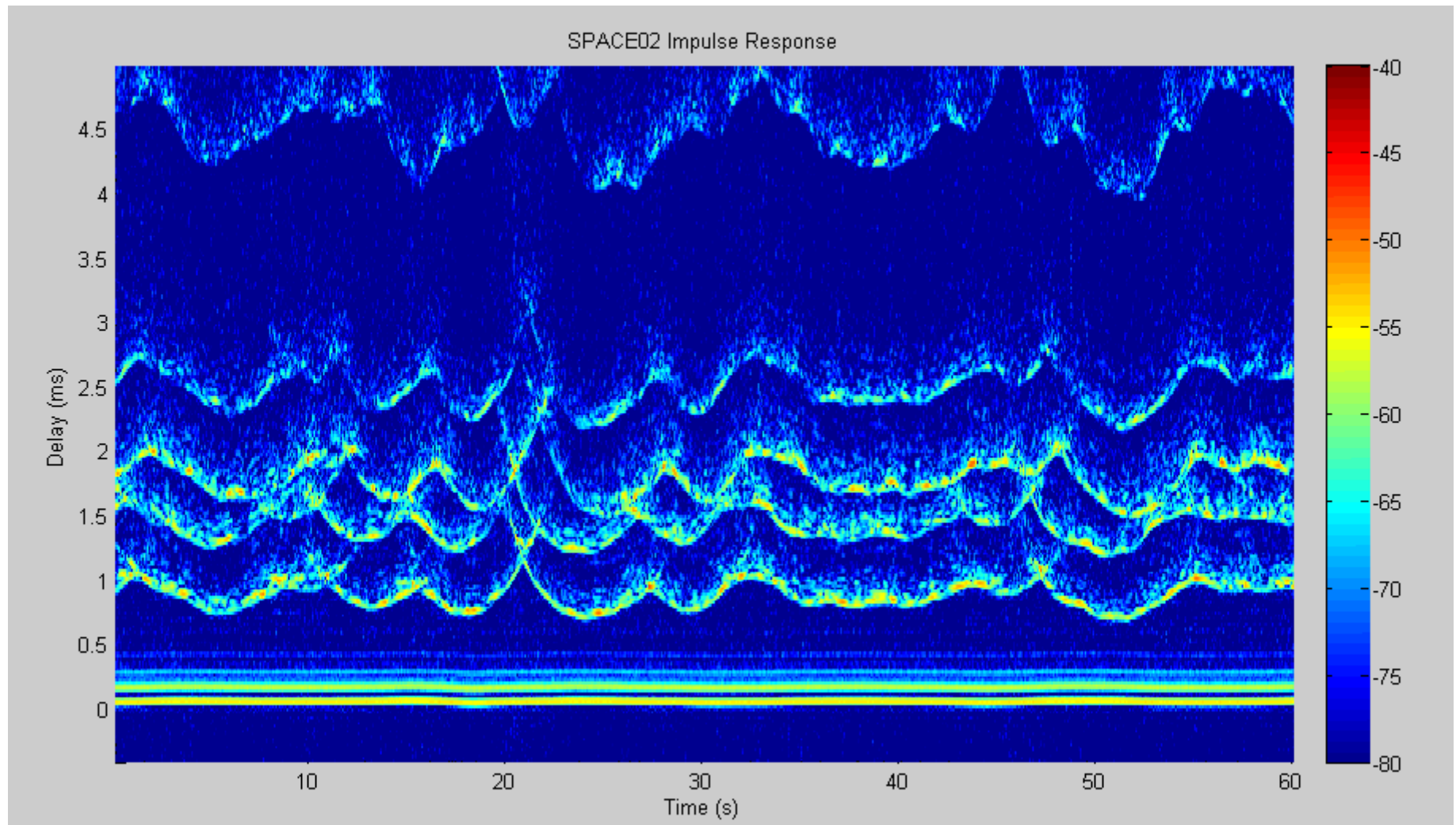


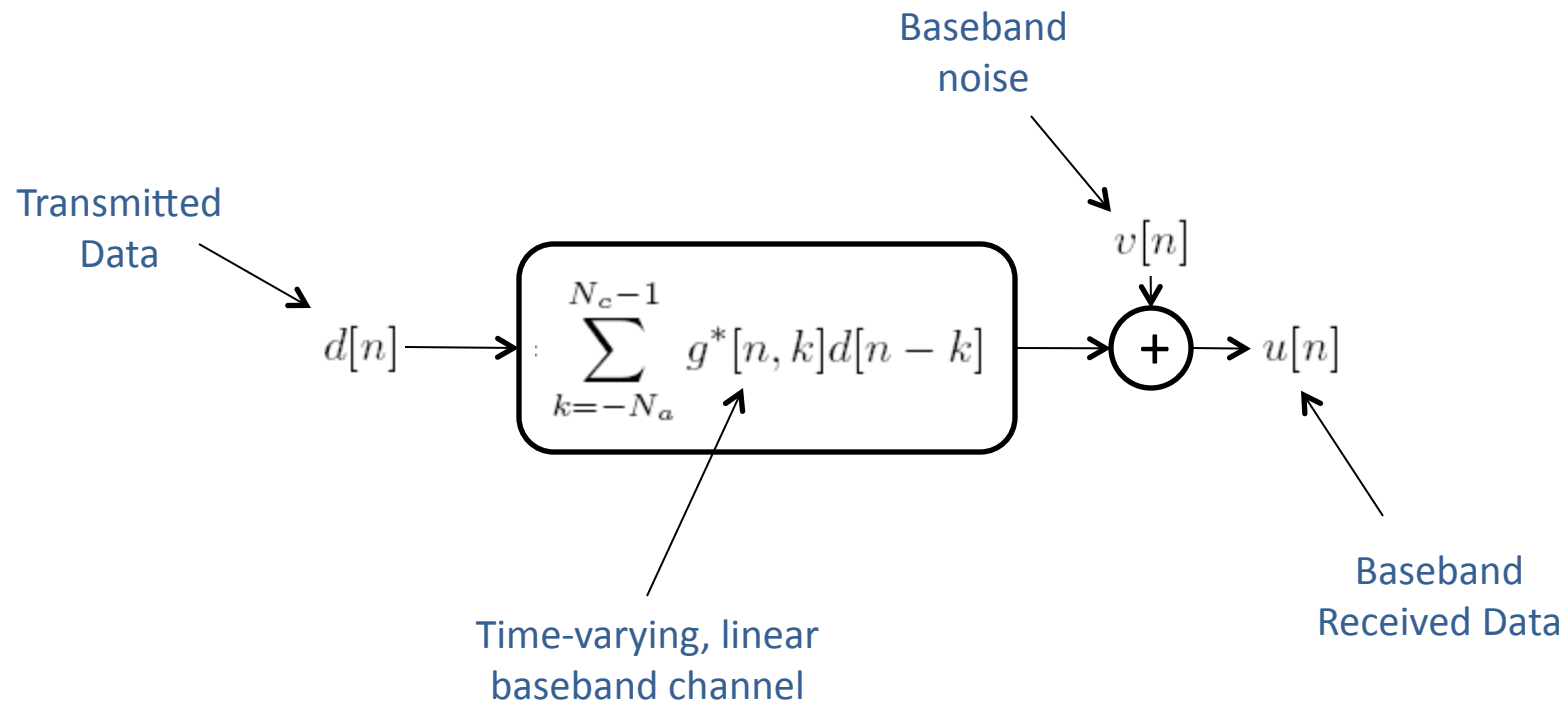
- Introduction:
 - Underwater Communication
 - Decision Feedback Equalization
 - Channel Estimate Based
 - Direct Adaptation
- Analysis of Equalization Behavior
- Simulation Results
- Summary and Conclusion
- Future Directions



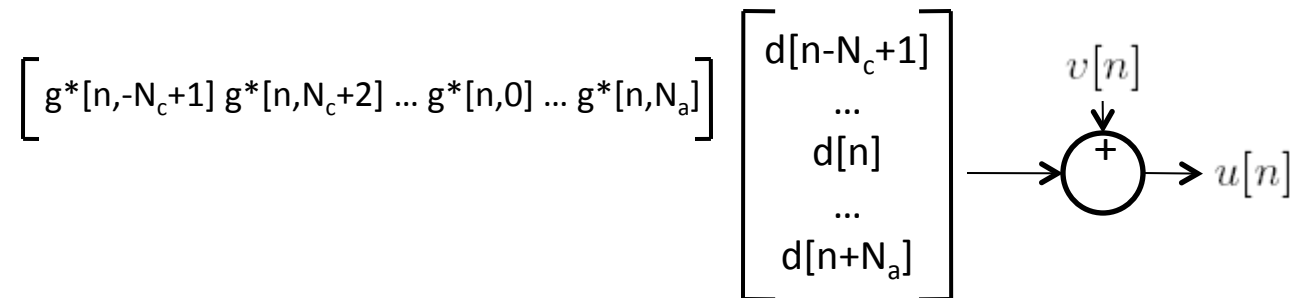


Time Varying Impulse Response

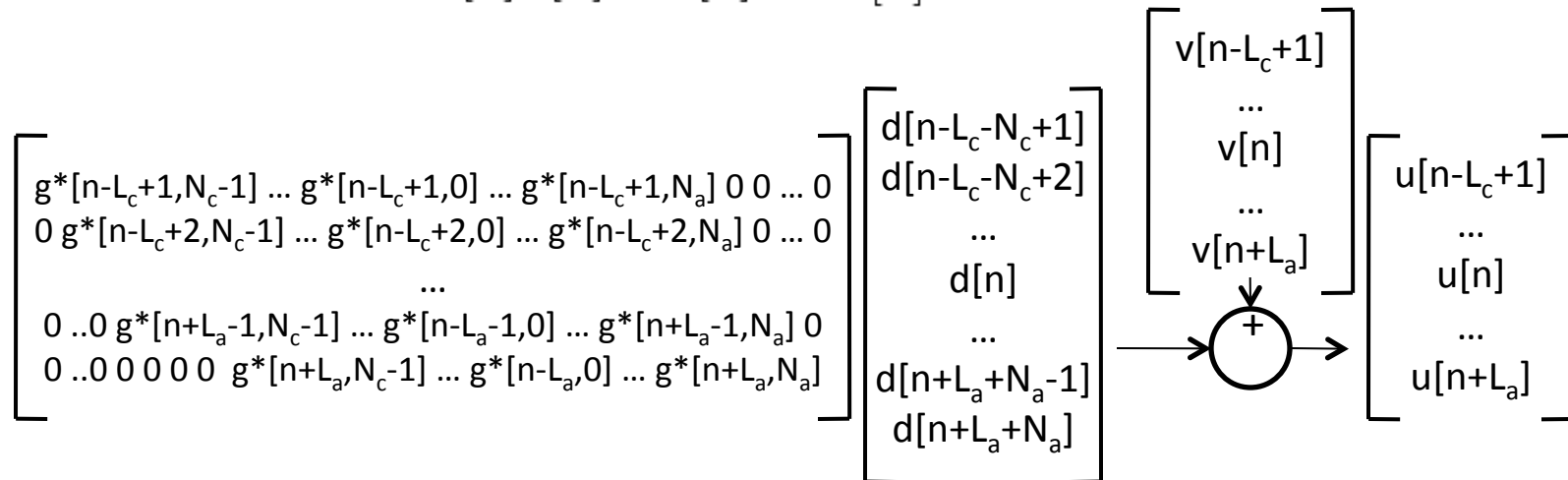




Vector-form: $\mathbf{g}^H[n] \mathbf{d}[n] + v[n] = u[n]$



Matrix Vector-form: $\mathbf{G}[n] \mathbf{d}[n] + \mathbf{v}[n] = \mathbf{u}[n]$



TX Data bit (linear) estimator: $\hat{d}[n] = \mathbf{h}^H [n] \mathbf{z}[n]$

↙ Vector of RX data and TX data estimates

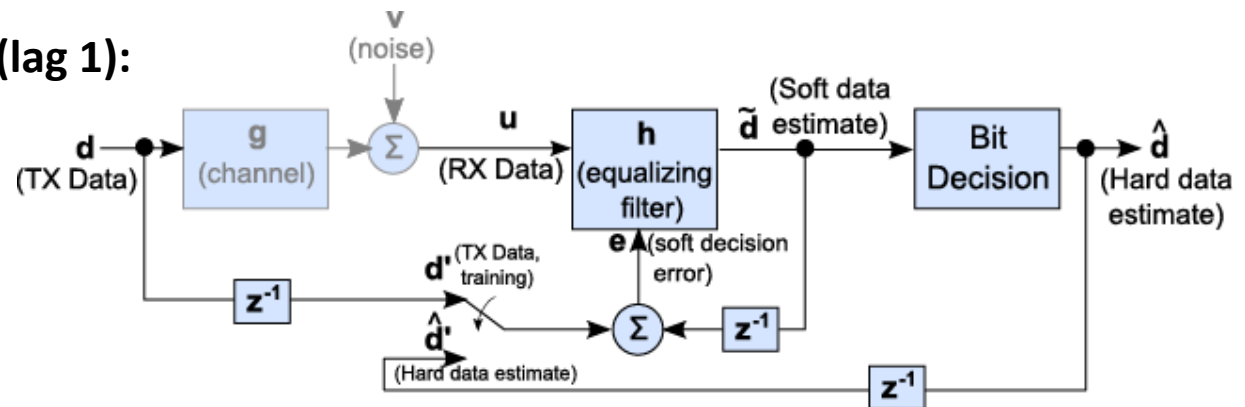
LMMSE Optimization: $\hat{\mathbf{h}}_{\text{opt}} = \arg \min_{\mathbf{h}'} E\{|\mathbf{h}^H \mathbf{z} - d|^2\}$

Solution: $\hat{\mathbf{h}}_{\text{opt}}[n] = \mathbf{R}_{\mathbf{z}}^{-1}[n] \mathbf{r}_{\mathbf{z}d}[n]$

$$\mathbf{R}_{\mathbf{z}}[n] = E\{\mathbf{z}\mathbf{z}^H\}$$

$$\mathbf{r}_{\mathbf{z}d} = E\{\mathbf{z}d^*\}$$

Recursive Processing (lag 1):





Decision Feedback Equalizer (DFE)



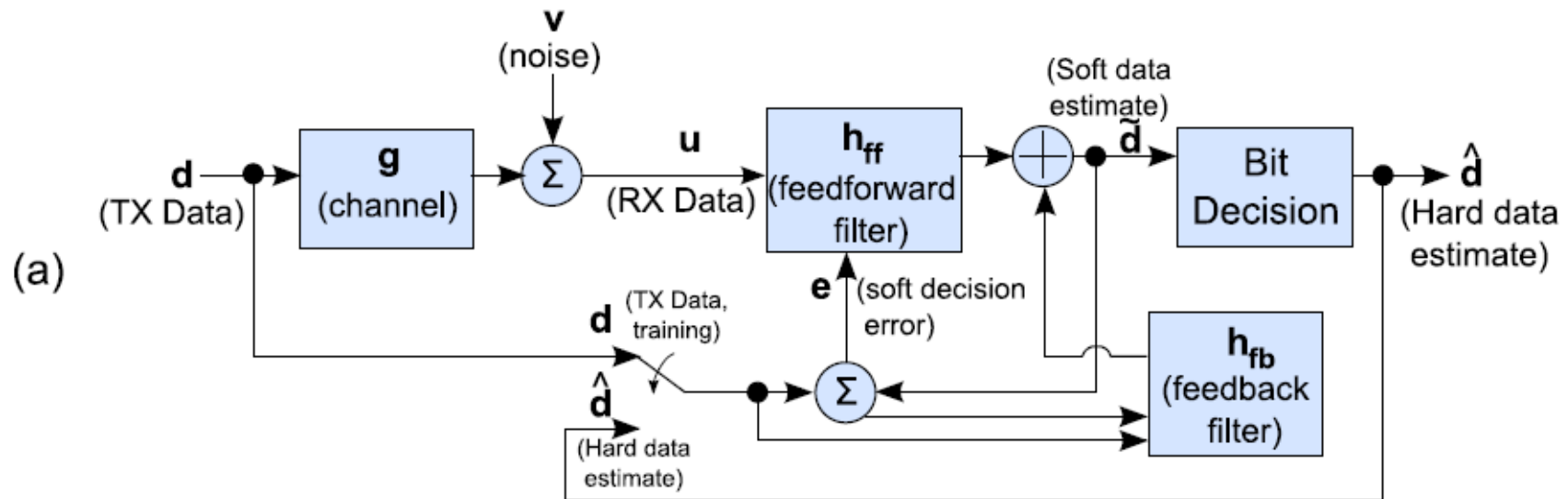
- Two Parts:
 - (Linear) feed-forward filter (of RX data)
 - (Linear) feedback filter (of data estimates)
- Estimate using RX data and TX data estimates

$$\mathbf{z}[n] = [u[n - L_c + 1] \dots u[n] \dots u[n + L_a], \hat{d}[n - 1] \dots \hat{d}[n - L_{fb}]]^T$$

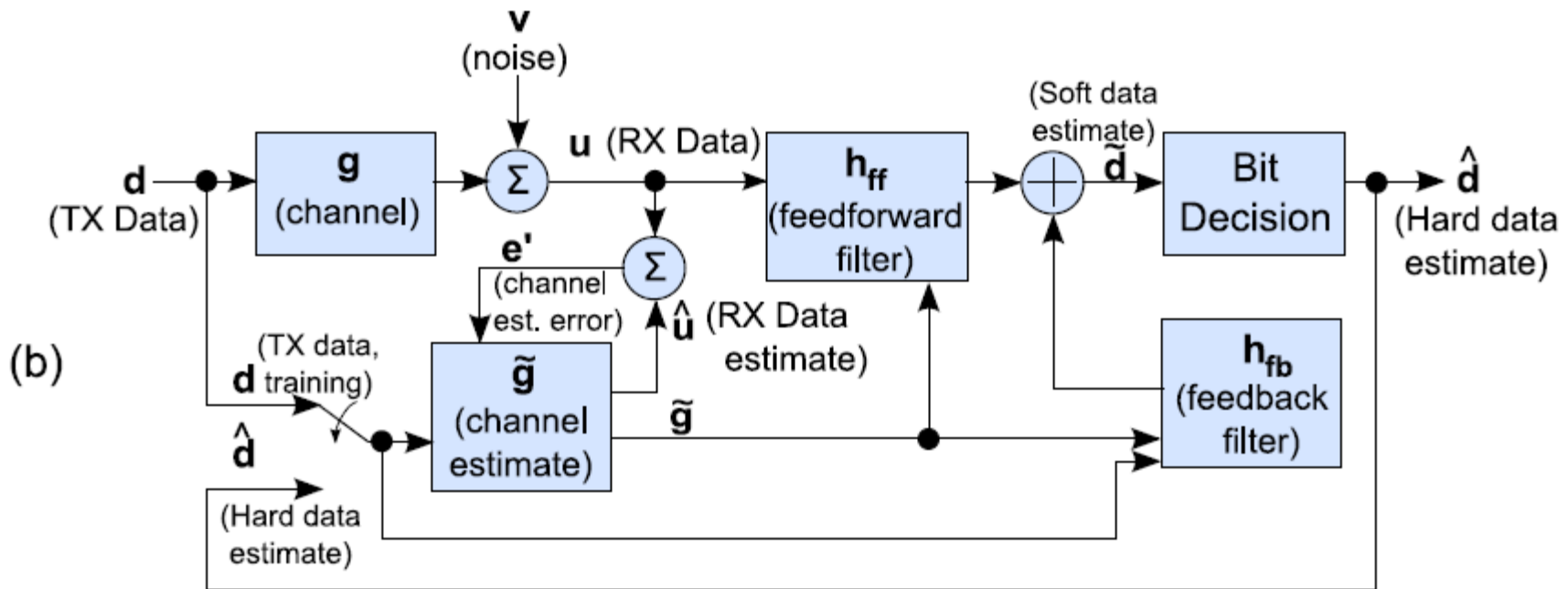
- Split Channel convolution Matrix: $\mathbf{G} \begin{matrix} \rightarrow \mathbf{G}_0 \\ \rightarrow \mathbf{G}_{fb} \end{matrix}$
 - Received data becomes: $\mathbf{u} = \mathbf{G}_0 \mathbf{d}_0 + \mathbf{G}_{fb} \mathbf{d}_{fb} + \mathbf{v}$

- Minimum Achievable Error:

$$\sigma_{0,dfc}^2 = 1 - \mathbf{g}_0^H [\mathbf{G}_0 \mathbf{G}_0^H + \mathbf{R}_v]^{-1} \mathbf{g}_0$$



$$\mathbf{h} = \begin{bmatrix} \mathbf{h}_{ff}[n] \\ \mathbf{h}_{fb}[n] \end{bmatrix} = \mathbf{E}\{\mathbf{z}\mathbf{z}^H\}^{-1} \mathbf{E}\{\mathbf{z}d^*\}$$



$$h_{ff} = [G_0 G_0^H + R_v]^{-1} G_s$$

$$h_{fb} = -G_{fb}^H h_{ff}$$

- Unit variance, white transmit data

$$E\{\mathbf{d}[n]\mathbf{d}^H[n]\} = \mathbf{I}$$

- TX data and obs. noise are uncorrelated

$$E\{\mathbf{v}[n]\mathbf{d}^H[m]\} = \mathbf{0}$$

- Obs. Noise variance:

$$\mathbf{R}_v = E\{\mathbf{v}[n]\mathbf{v}^H[n]\}$$

- Perfect data estimation (for feedback)

$$\hat{\mathbf{d}} = \mathbf{d}$$

- Equalizer Length = Estimated Channel Length

$$N_a + N_c = L_a + L_c$$

- MMSE Equalizer Coefficients have form:

$$\mathbf{h}_{ff} = [\mathbf{G}_0 \mathbf{G}_0^H + \mathbf{R}_v]^{-1} \mathbf{G}_0 \mathbf{s}$$

$$\mathbf{h}_{fb} = -\mathbf{G}_{fb}^H \mathbf{h}_{ff}$$



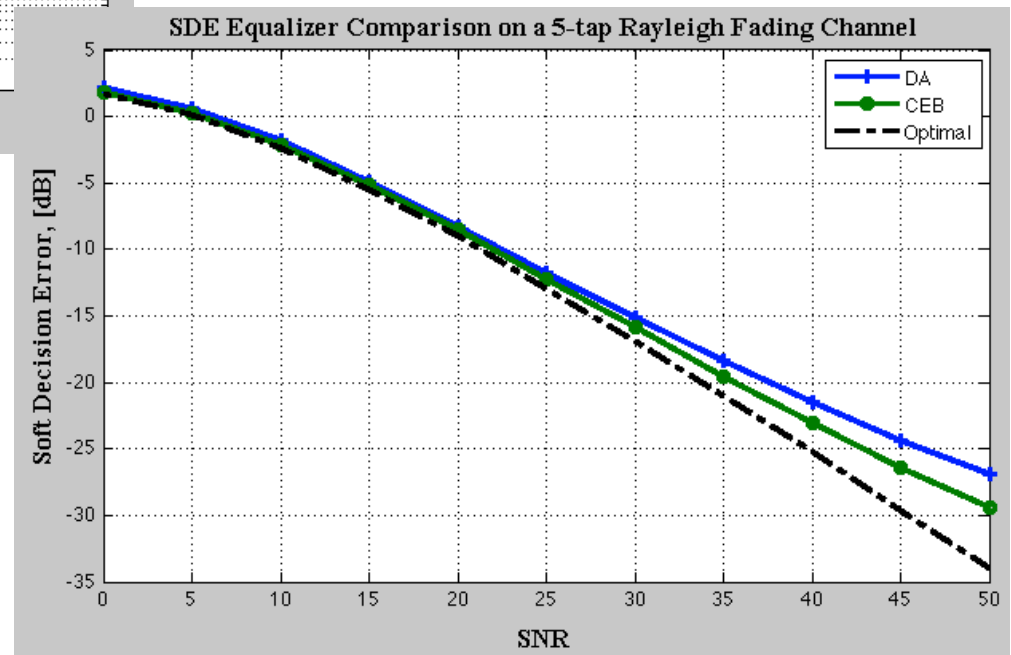
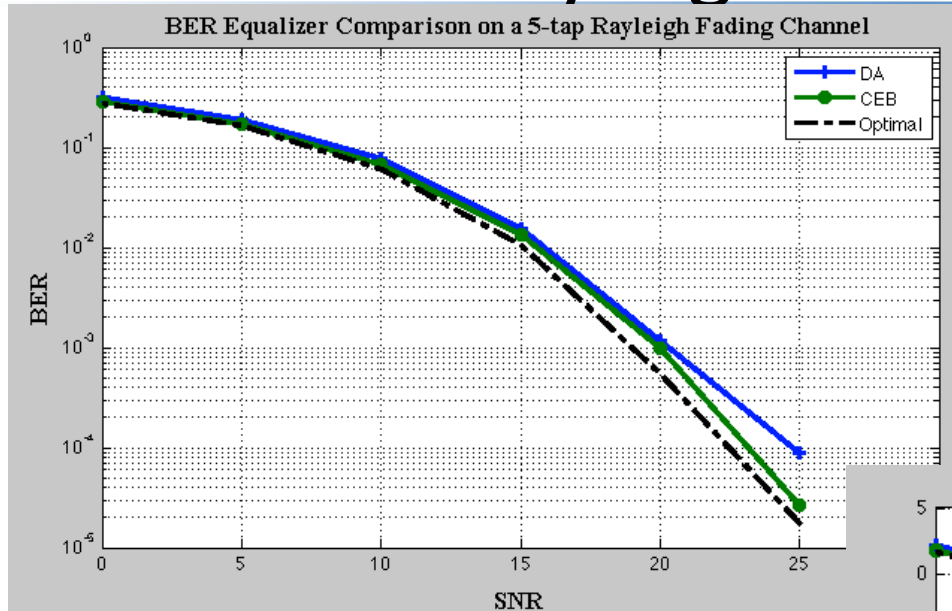
Comparison between DA and CEB



- In the past, CEB methods empirically shown to have lower mean squared error at high SNR
- Reasons for difference varied:
 - Condition number of correlation matrix
 - Number of samples required to get good est.
- Analysis to follow: low and high SNR regimes



Comparison of DA and CEB on Rayleigh Fading Channel



Why the difference?

- Correlation time
 - DA equalizer taps have lower correlation time at high SNR
 - At low SNR, two methods equivalent
- But how do we show this?
 - Combination of analysis and simulation

- Simple channel model to analyze
- Similar to encountered situations

$$g[n + 1] = \alpha g[n] + w[n]$$

$$R_{gg}[k] = E\{g[n]g^*[n + k]\} = \begin{cases} \sigma_w^2 \left(\frac{(\alpha^*)^k}{1 - |\alpha|^2} \right) & k \geq 0 \\ \sigma_w^2 \left(\frac{\alpha^{-k}}{1 - |\alpha|^2} \right) & k < 0 \end{cases}$$

$$\begin{aligned}\mathbf{h}_{\text{ff}}[n+1] &= (\mathbf{G}_0[n+1]\mathbf{G}_0^H[n+1] + \mathbf{R}_v)^{-1}(\mathbf{g}_0[n+1]) \\ &\approx \mathbf{R}_v^{-1}(\alpha\mathbf{g}_0[n] + \mathbf{w}[n]) \\ &\approx \alpha\mathbf{h}_{\text{ff}}[n] + \mathbf{R}_v^{-1}\mathbf{w}[n]\end{aligned}$$

$$\mathbf{R}_v + \mathbf{G}[n]\mathbf{G}^H[n] \approx \mathbf{R}_v$$

$$\mathbf{h}_{\text{ff}}[n + 1] = (\mathbf{G}_0[n + 1]\mathbf{G}_0^H[n + 1] + \mathbf{R}_v)^{-1}(\mathbf{g}_0[n + 1])$$

$$\mathbf{Q}[n] = \mathbf{G}_0[n]\mathbf{G}_0^H[n] + \mathbf{R}_v \approx \mathbf{G}_0[n]\mathbf{G}_0^H[n]$$

$$(\mathbf{G}_0[n]\mathbf{G}_0^H[n])\mathbf{h}_{\text{ff}}[n] = \mathbf{g}_0[n]$$

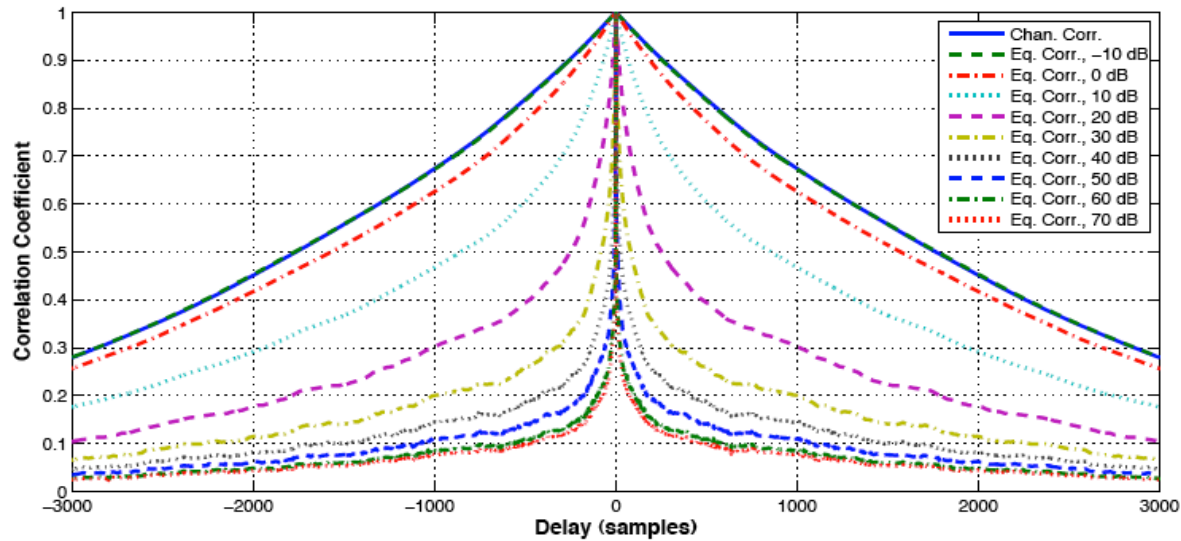
$$(\mathbf{G}_0[n] \mathbf{G}_0^H[n]) \mathbf{h}_{\text{ff}}[n] = \mathbf{g}_0[n]$$

$$\mathbf{G}_0 = \begin{bmatrix} g_0^*[n-L+1] & 0 & 0 & \cdots & 0 \\ g_1^*[n-L+2] & g_0^*[n-L+2] & 0 & \cdots & 0 \\ g_2^*[n-L+3] & g_1^*[n-L+3] & g_0^*[n-L+3] & \cdots & 0 \\ \vdots & & & \ddots & \vdots \\ g_L^*[n] & g_{L-1}^*[n] & g_{L-2}^*[n] & \cdots & g_0^*[n] \end{bmatrix}$$

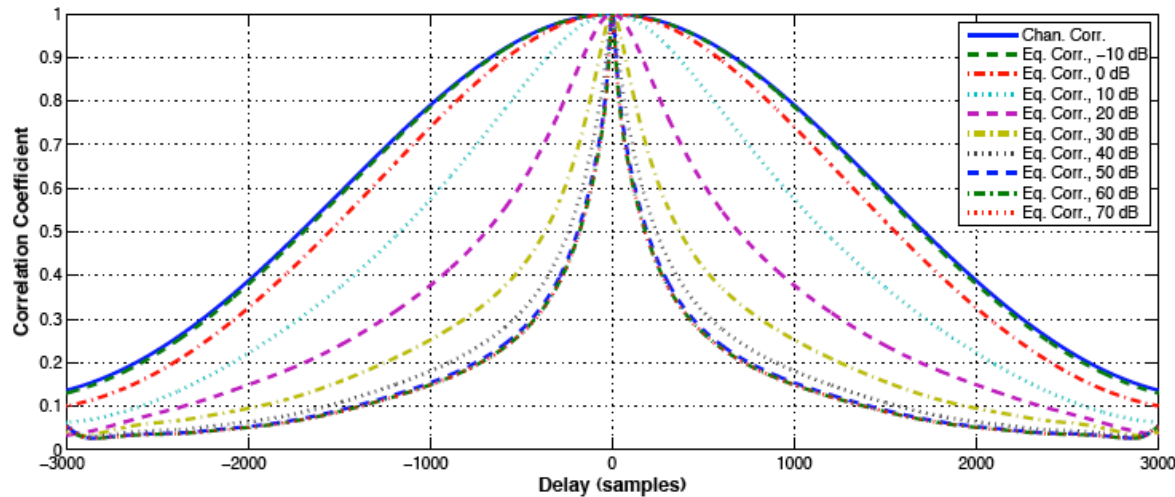
$$\mathbf{G}_0 \mathbf{G}_0^H = \begin{bmatrix} |g_0|^2 & \cdots \\ g_0 g_1^* & \cdots \\ g_0 g_2^* & \cdots \\ \vdots & \vdots \\ g_0 g_L & \cdots \end{bmatrix} \quad \mathbf{h}[n] = \begin{bmatrix} 1/g_0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Correlation over SNR

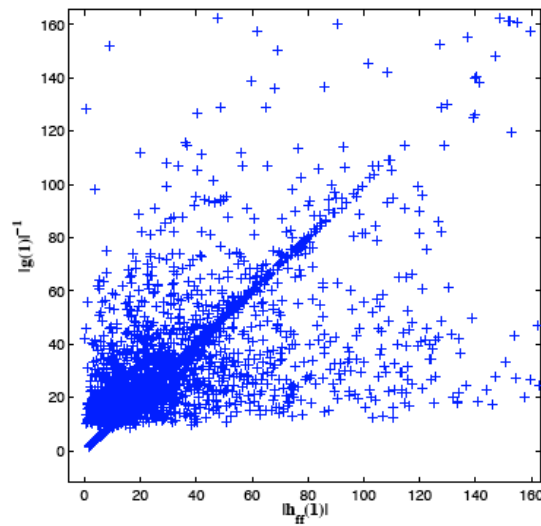
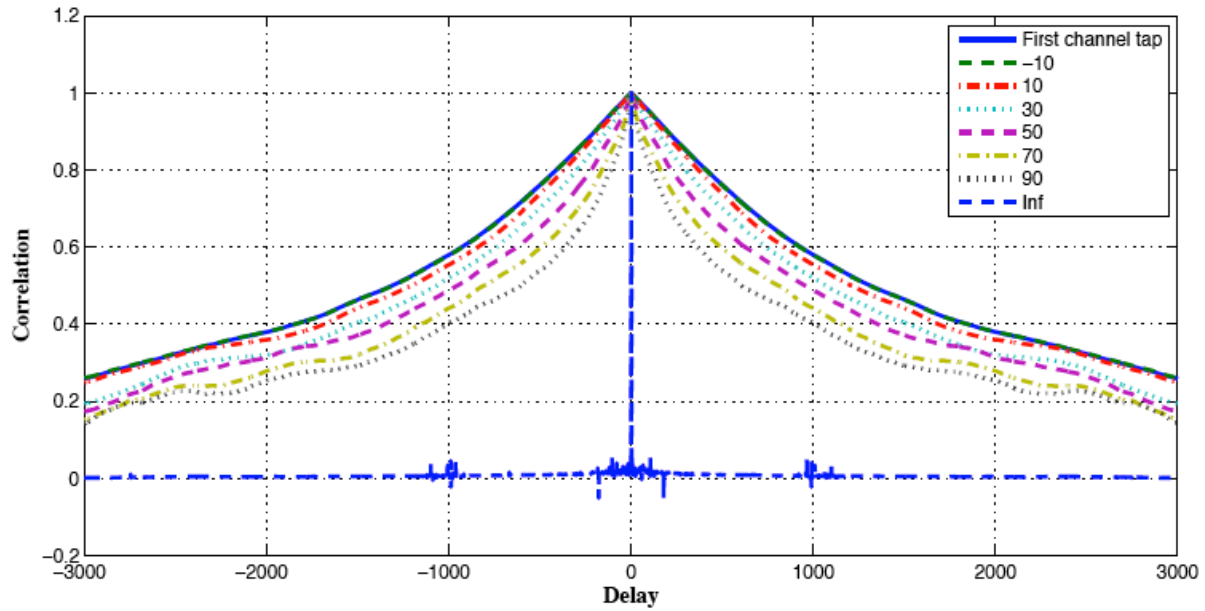
AR(1)
model



Gaussian
model



Multi-tap
AR(1)
model



- As SNR increases, correlation time of equalizer taps is reduced
 - CEB is tracking value correlated over longer time
 - DA should do worse
- Assumed noise statistics were stationary
 - Not always case in underwater
- Underwater communication is power limited
 - Operate in low SNR regime (<35dB)

- Include channel state information into DA
 - Sparsity
- Reduce number of snapshots for channel model
 - Physical constraints?
 - Compressed sensing?

Questions?

