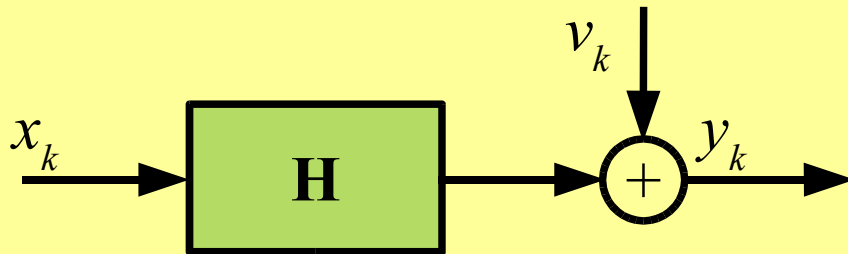


Structure and Design of MMSE Channel Equalizers

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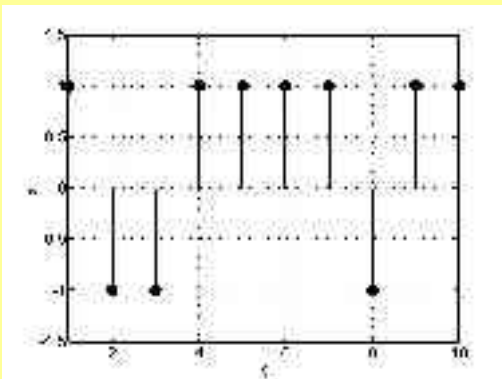
www.ee.hacettepe.edu.tr/~toker/equalizers.pdf

Intersymbol Interference (ISI)

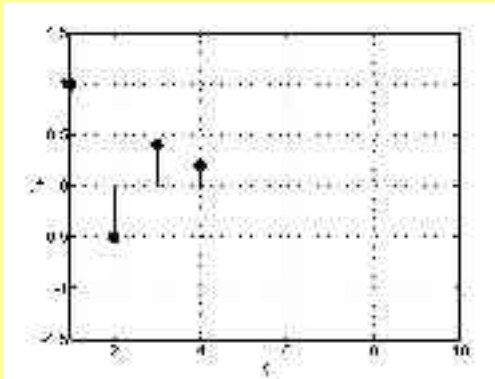


$$y_k = \sum_{n=0}^{N_H-1} h_n x_{k-n} + v_k$$

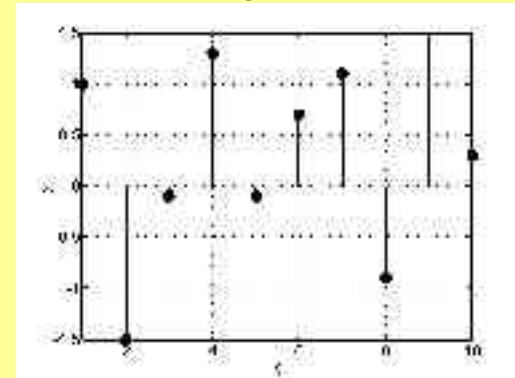
transmitted data



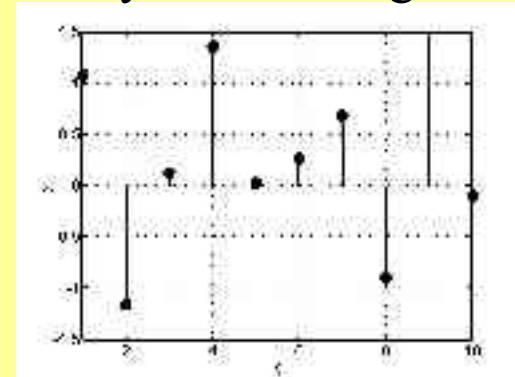
ISI channel



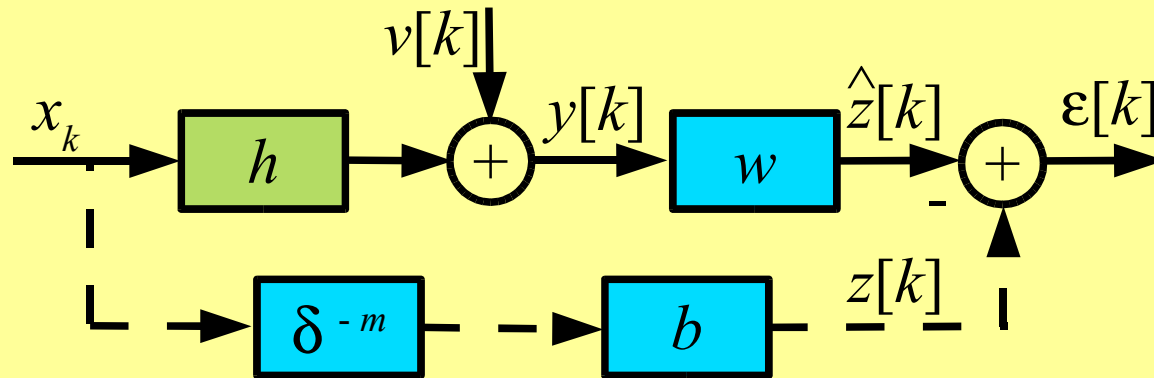
received signal-no noise



noisy received signal

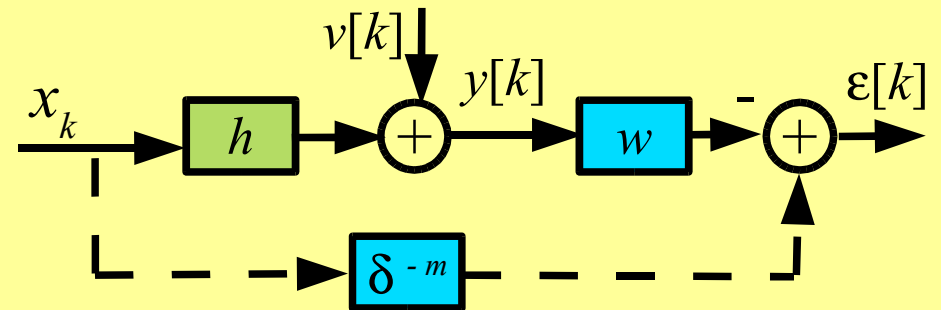


MMSE FIR Equalizers



- *Equalizer:*

$$b = 1$$

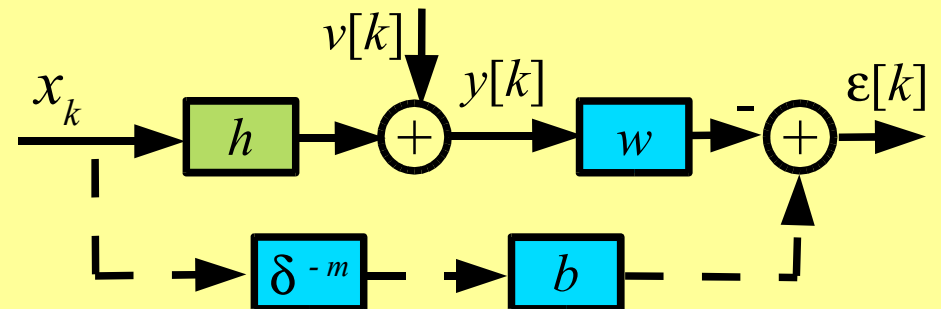


- *Partial response equalizer:*

b: fixed (for example, $b=1+z^{-1}$)

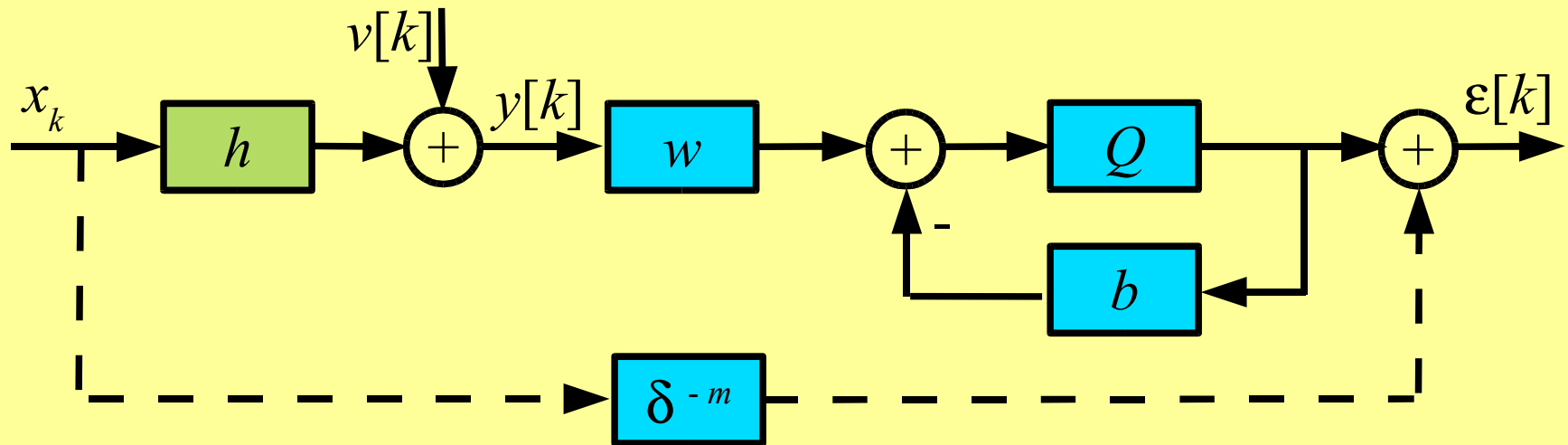
- *Channel shortening equalizer:*

b: design parameter

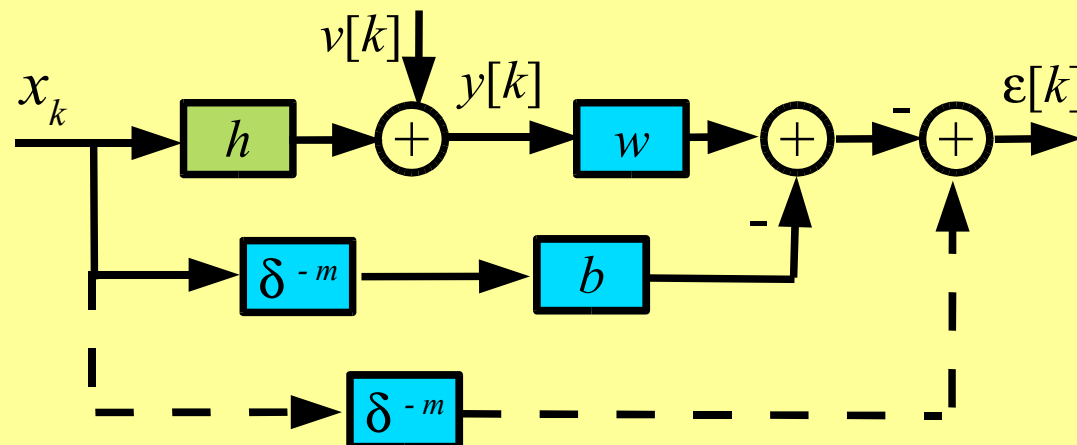


MMSE FIR

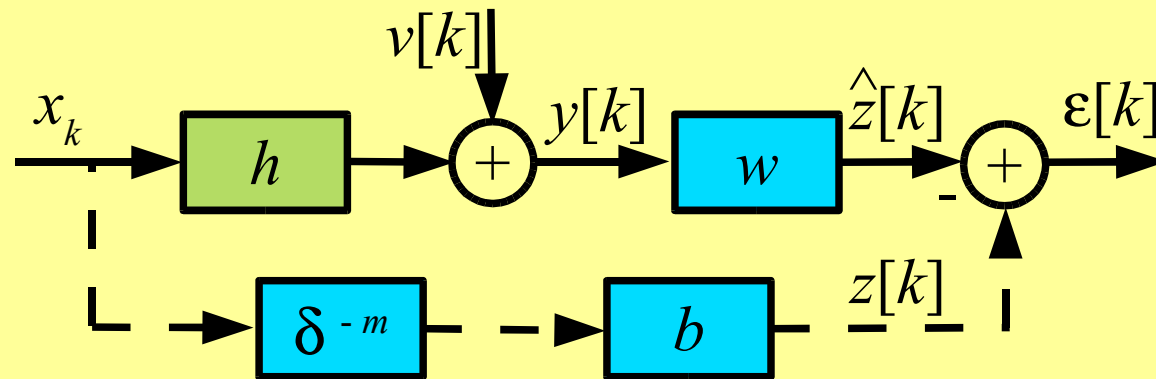
Decision Feedback Equalizer (DFE)



- Assumption: *Past decisions are correct*



MMSE FIR Equalizers



$$\hat{z}_k = \sum_{n=0}^{n_w-1} w_n^* y_{k-n} = \begin{bmatrix} w_0^* & w_1^* & \cdots & w_{n_w-1}^* \end{bmatrix} \begin{bmatrix} y_k \\ y_{k-1} \\ \vdots \\ y_{k-(n_w-1)} \end{bmatrix} = \mathbf{w}^H \mathbf{y}_k$$

$$y_k = \sum_{m=0}^{n_w-1} w_m^* \left(\sum_{n=0}^{n_h-1} h_n x_{k-n-m} + v_{k-m} \right)$$

$$\hat{z}_k = \begin{bmatrix} w_0^* & w_1^* & \cdots & w_{n_w-1}^* \end{bmatrix} \left(\begin{bmatrix} h_0 & h_1 & \cdots & h_{n_h-1} & 0 & \cdots & 0 \\ 0 & h_0 & h_1 & \cdots & h_{n_h-1} & \cdots & 0 \\ 0 & \ddots & \ddots & \ddots & \cdots & \ddots & 0 \\ 0 & \cdots & 0 & h_0 & h_1 & \cdots & h_{n_h-1} \end{bmatrix} \begin{bmatrix} x_k \\ x_{k-1} \\ \vdots \\ x_{k-(n_h+n_w-2)} \end{bmatrix} + \begin{bmatrix} v_k \\ v_{k-1} \\ \vdots \\ v_{k-(n_w-1)} \end{bmatrix} \right)$$

$$\hat{z}_k = \mathbf{w}^H \mathbf{H} \mathbf{x}_k + \mathbf{w}^H \mathbf{v}_k$$

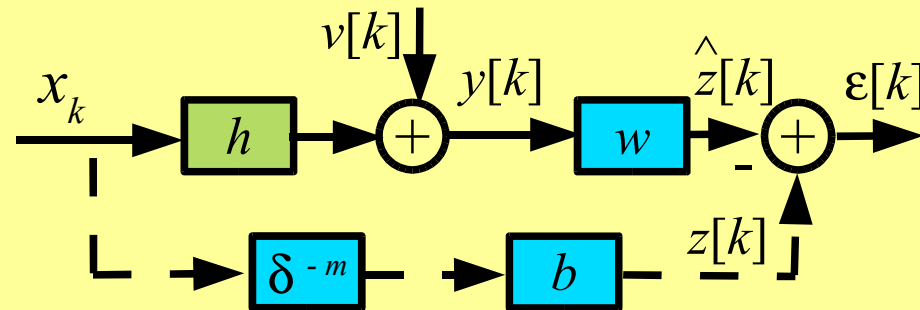
$$z_k = \sum_{n=0}^{n_b-1} b_{n-m}^* x_{k-n}$$

$$= \begin{bmatrix} 0 & \cdots & 0 & b_0^* & b_1^* & \cdots & b_{n_b-1}^* & 0 & \cdots & 0 \end{bmatrix} \begin{bmatrix} x_k \\ x_{k-1} \\ \vdots \\ x_{k-(n_w+n_h-2)} \end{bmatrix}$$

$$= \begin{bmatrix} \mathbf{0}_{m \times 1} \\ \mathbf{b} \\ \mathbf{0} \end{bmatrix}^H \mathbf{x}_k$$

$$z_k = \mathbf{b}^H \mathbf{x}_k$$

MMSE FIR Equalizers



- Error:

$$\begin{aligned}\varepsilon_k &= z_k - \hat{z}_k \\ &= [\tilde{\mathbf{b}}^H - \mathbf{w}^H \mathbf{H}] \mathbf{x}_k - \mathbf{w}^H \mathbf{v}_k\end{aligned}$$

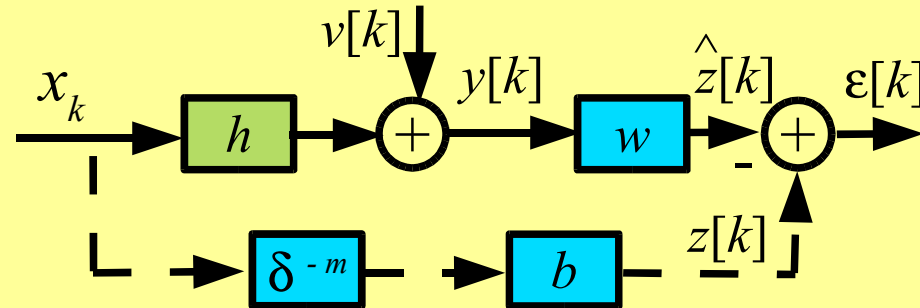
- Mean Square Error (MSE):

$$\begin{aligned}J &= E\{\varepsilon_k \varepsilon_k^*\} \\ &= [\tilde{\mathbf{b}}^H - \mathbf{w}^H \mathbf{H}] \mathbf{R}_{xx} [\tilde{\mathbf{b}} - \mathbf{H}^H \mathbf{w}] + \mathbf{w}^H \mathbf{R}_{vv} \mathbf{w}\end{aligned}$$

$$\mathbf{R}_{xx} = E\{\mathbf{x}_k \mathbf{x}_k^H\}$$

$$= E\left\{ \begin{bmatrix} x_k \\ x_{k-1} \\ \vdots \\ x_{k-(n_h+n_w-2)} \end{bmatrix} \begin{bmatrix} x_k^* & x_{k-1}^* & \cdots & x_{k-(n_h+n_w-2)}^* \end{bmatrix} \right\} = \begin{bmatrix} r_{xx}[0] & r_{xx}[1] & \cdots & r_{xx}[n_h+n_w-1] \\ r_{xx}^*[-1] & r_{xx}[0] & \cdots & r_{xx}[n_h+n_w-2] \\ \vdots & \vdots & \ddots & \vdots \\ r_{xx}^*[-(n_h+n_w-1)] & r_{xx}^*[-(n_h+n_w-2)] & \cdots & r_{xx}[0] \end{bmatrix}$$

MMSE FIR Equalizers



- J : quadratic function; optimum w is found by differentiating J wrt w and equating to zero^[4] (or by orthogonality principle, $E\{\epsilon[k]y^H[k]\}=0$ ^[1,2]).

$$\begin{aligned} w^H &= \tilde{b}^H R_{xx} H^H (H R_{xx} H^H + R_{vv})^{-1} \\ &= \tilde{b}^H (R_{xx}^{-1} + H^H R_{vv}^{-1} H)^{-1} H^H R_{vv}^{-1} \quad (*) \end{aligned}$$

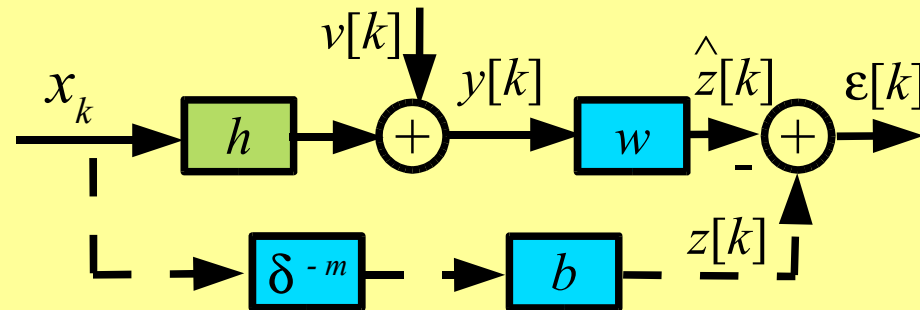
- MMSE FIR equalizer, $b=1$: the optimum receiver is

$$w^H = e_m^H (R_{xx}^{-1} + H^H R_{vv}^{-1} H)^{-1} \underbrace{H^H R_{vv}^{-1}}_{\text{whitening matched filter}}$$

$$e_m = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \leftarrow m$$

(*): Matrix Inversion Lemma: $(R_{xx}^{-1} + H^H R_{vv}^{-1} H)^{-1} = R_{xx} - R_{xx} H^H (H R_{xx} H^H + R_{vv})^{-1} H R_{xx}$

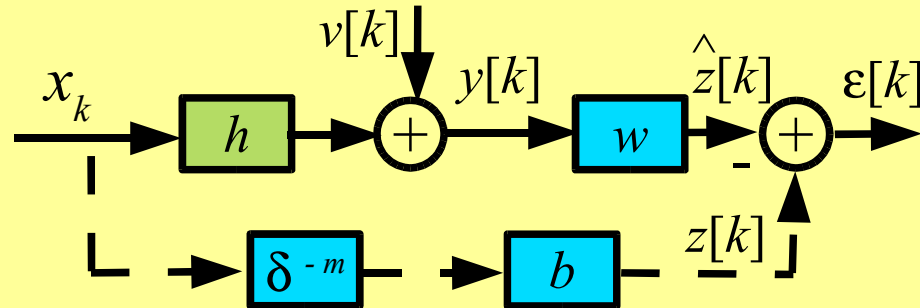
MMSE FIR Partial Response Equalizers



- MMSE FIR fixed partial response equalizer, \mathbf{b} fixed:
- the optimum receiver is

$$\mathbf{w}^H = \tilde{\mathbf{b}}^H \left(\mathbf{R}_{xx}^{-1} + \mathbf{H}^H \mathbf{R}_{vv}^{-1} \mathbf{H} \right)^{-1} \underbrace{\mathbf{H}^H \mathbf{R}_{vv}^{-1}}_{\text{whitening matched filter}}$$

MMSE FIR Channel Shortening Equalizers



- MMSE FIR channel shortening equalizer, \mathbf{b} fixed length variable filter.

- Substitute \mathbf{w} into J :

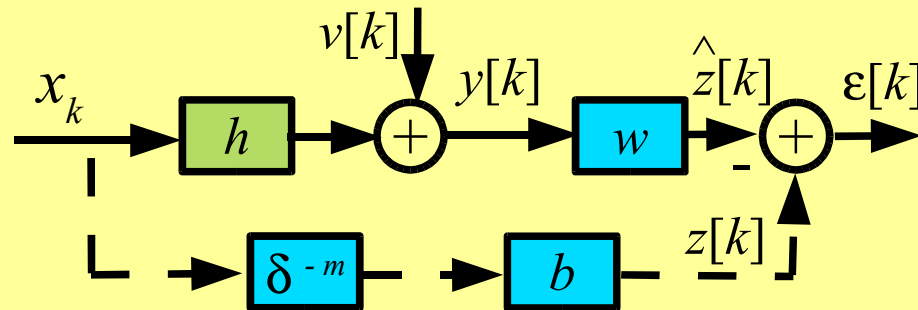
$$J = \tilde{\mathbf{b}}^H \left(\mathbf{R}_{xx}^{-1} + \mathbf{H}^H \mathbf{R}_{vv}^{-1} \mathbf{H} \right)^{-1} \tilde{\mathbf{b}}$$

$$= \begin{bmatrix} 0 & \mathbf{b}^H & 0 \end{bmatrix} \begin{bmatrix} & & & 0 \\ & \mathbf{R} & & \mathbf{b} \\ & & & 0 \end{bmatrix}$$

$$= \mathbf{b}^H \mathbf{R} \mathbf{b}$$

- Impose a constraint on \mathbf{b} in order to avoid the trivial case $\mathbf{b}=\mathbf{0}$, $J=0$
 - Orthogonality constraint $\mathbf{b}^H \mathbf{b} = 1$.

MMSE FIR Channel Shortening Equalizers



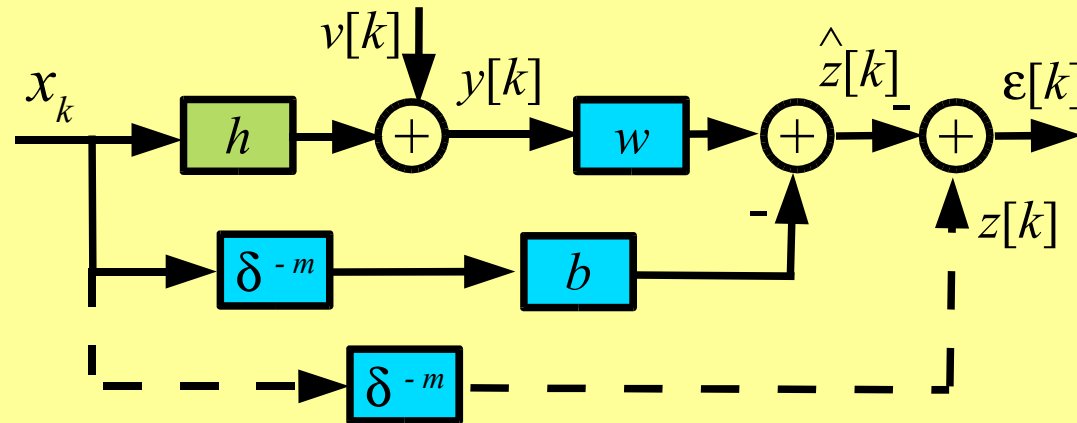
- Problem becomes :

$$\begin{aligned}
 \text{minimize } J &= \mathbf{b}^H \mathbf{R} \mathbf{b} \\
 &= \mathbf{b}^H \mathbf{U} \Delta \mathbf{U}^H \mathbf{b} \\
 &= \mathbf{b}^H \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_{n_b} \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_{n_b} \end{bmatrix} \begin{bmatrix} \mathbf{u}_1^H \\ \mathbf{u}_2^H \\ \vdots \\ \mathbf{u}_{n_b}^H \end{bmatrix} \mathbf{b} \\
 \text{subject to } & \mathbf{b}^H \mathbf{b} = 1
 \end{aligned}$$

- Constrained optimization: Solution for \mathbf{b} is the eigenvector corresponding to the smallest eigenvalue, λ_{min} , of \mathbf{R} .

$$MMSE = \lambda_{min}$$

MMSE FIR Decision Feedback Equalizer (DFE)

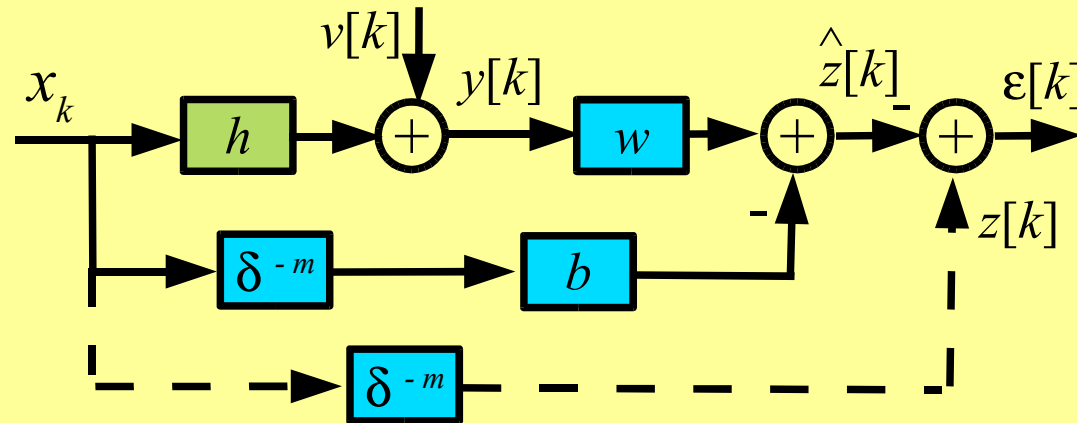


$$\hat{z}_k = \sum_{n=0}^{n_w-1} w_n y_{k-n} - \sum_{\textcircled{n=1}}^{n_b-1} b_n x_{k-n-m} \quad \longleftarrow \quad y[k] = \sum_{n=0}^{n_h-1} h[n] x[k-n] + v[k]$$

$$= \sum_{n=0}^{n_h+n_w-1} c_n x_{k-n} - \sum_{n=1}^{n_b-1} b_n x_{k-n-m} + \sum_{n=0}^{n_w-1} w_n v_{k-n} \quad \longleftarrow \quad c[n] = w[n] * h[n]$$

$$= \underbrace{c_m x_{k-m}}_{\text{information bearing cursor}} + \underbrace{\sum_{n=0}^{m-1} c_n x_{k-n}}_{\text{residual precursor ISI}} + \underbrace{\sum_{n=1}^{n_b-1} (c_{n+m} - d_n) x_{k-n-m}}_{\text{modeled postcursor ISI}} + \underbrace{\sum_{n=m+n_b+1}^{n_h+n_w-1} c_n x_{k-n}}_{\text{residual postcursor ISI}} + \underbrace{\sum_{n=0}^{n_w-1} w_n v_{k-n}}_{\text{filtered noise}}$$

MMSE FIR Decision Feedback Equalizer (DFE)



$$\hat{z}_k = \underbrace{c_m x_{k-m}}_{\text{information bearing cursor}} + \underbrace{\sum_{n=0}^{m-1} c_n x_{k-n}}_{\text{residual precursor ISI}} + \underbrace{\sum_{n=1}^{n_b-1} (c_{n+m} - b_n) x_{k-n-m}}_{\text{modeled postcursor ISI}} + \underbrace{\sum_{n=m+n_b+1}^{n_n+n_w-1} c_n x_{k-n}}_{\text{residual postcursor ISI}} + \underbrace{\sum_{n=0}^{n_w-1} w_n v_{k-n}}_{\text{filtered noise}}$$

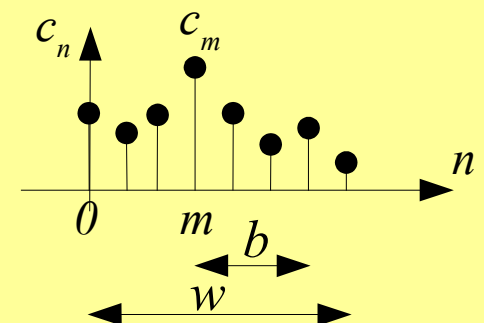
- Goals:

- Feedforward filter:

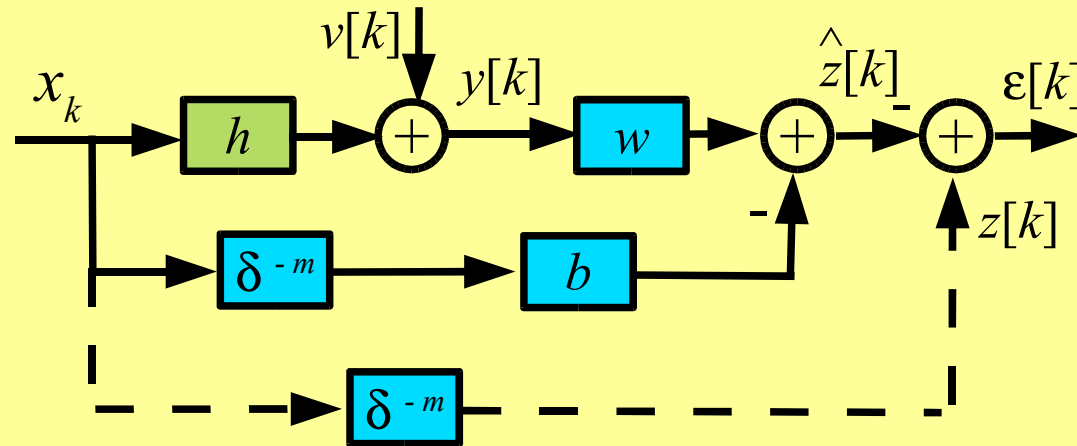
- shape $c_n = h_n * w_n$ so that $c_m \approx 1$
- small residual ISI, $c_n \approx 0, n \neq m$
- keep noise gain $\sum_k |f_k|^2$ as small as possible,

- Feedback filter:

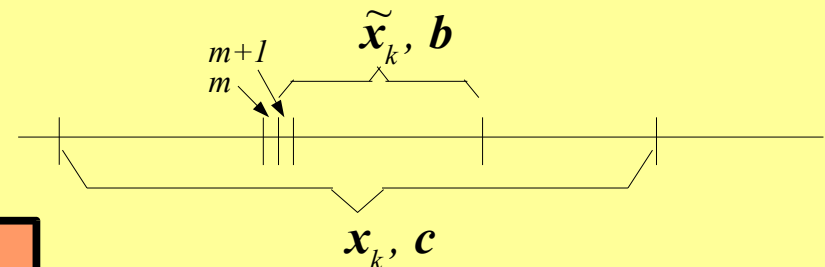
- cancel the remaining ISI by matching $d_n = c_n + m, n = 1, \dots, n_b - 1$



MMSE FIR Decision Feedback Equalizer (DFE)



$$\varepsilon_k = x_{k-m} - \{ \mathbf{w}^H (\mathbf{H}\mathbf{x}_k + \mathbf{v}_k) - \mathbf{b}^H \tilde{\mathbf{x}}_k \}$$



• Assumptions:

$$E \{ x_{k-m} \tilde{\mathbf{x}}_k \} = 0, \quad E \{ x_{k-m} \mathbf{v}_k \} = 0, \quad E \{ \mathbf{x}_k \mathbf{v}_k^H \} = 0$$

$$\tilde{\mathbf{x}}_k = \mathbf{M}\mathbf{x}_k$$

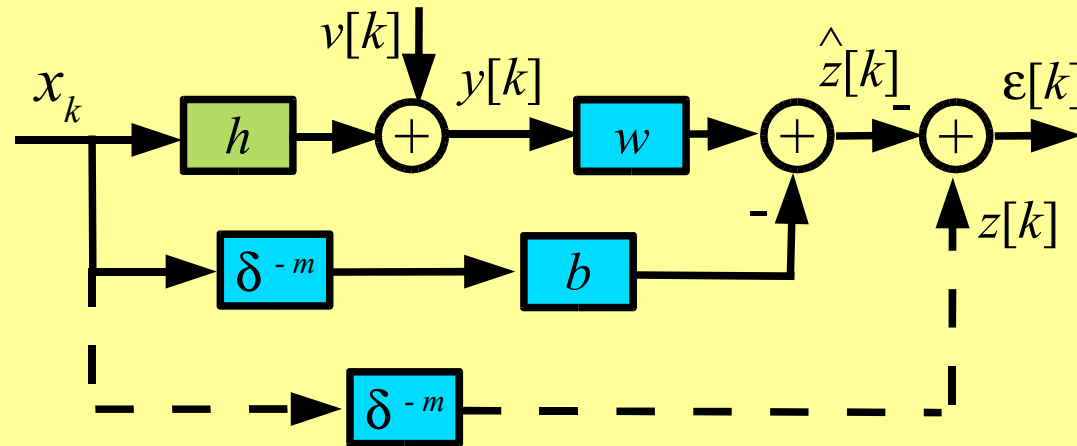
$$\mathbf{M} = \begin{bmatrix} \mathbf{0}_{n_b \times m} & \mathbf{I}_{n_b \times n_b} & \mathbf{0}_{n_b \times (n_h + n_w - n_b - m)} \end{bmatrix}$$

$$J = E \{ \varepsilon[k] \varepsilon^*[k] \}$$

$$= \sigma_x^2 - \left[\sigma_x^2 \mathbf{e}_m^H + \mathbf{b}^H \mathbf{R}_{xx} \right] \mathbf{H}^H \mathbf{w} - \mathbf{w}^H \mathbf{H} \left[\sigma_x^2 \mathbf{e}_m + \mathbf{R}_{xx} \mathbf{b} \right] + \mathbf{w}^H \left[\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv} \right] \mathbf{w} + \mathbf{b}^H \mathbf{R}_{xx} \mathbf{b}$$

MMSE FIR

Decision Feedback Equalizer (DFE)



$$J = \sigma_x^2 - \left[\sigma_x^2 \mathbf{e}_m^H + \mathbf{b}^H \mathbf{R}_{xx} \right] \mathbf{H}^H \mathbf{w} - \mathbf{w}^H \mathbf{H} \left[\sigma_x^2 \mathbf{e}_m + \mathbf{R}_{xx} \mathbf{b} \right] + \mathbf{w}^H \left[\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv} \right] \mathbf{w} + \mathbf{b}^H \mathbf{R}_{xx} \mathbf{b}$$

- Optimum feedback filter, $\nabla_{\mathbf{b}} J = 0$:

$$\mathbf{b}^H = \mathbf{w}^H \mathbf{H} \mathbf{M}^H$$

- Optimum feedforward filter, $J \leftarrow \mathbf{b}$, $\nabla_{\mathbf{w}} J = 0$:

$$\mathbf{w}^H = \sigma_x^2 \mathbf{e}_m^H \mathbf{H} \left(\mathbf{H} \left(\mathbf{I} - \mathbf{M}^H \mathbf{R}_{xx} \mathbf{M} \right) \mathbf{H}^H + \mathbf{R}_{vv} \right)^{-1}$$

- Equivalent optimum feedforward filter, $\nabla_{\mathbf{w}} J = 0$:

$$\mathbf{w}^H = \left(\sigma_x^2 \mathbf{e}_m^H + \mathbf{b}^H \mathbf{M} \mathbf{R}_{xx} \right) \mathbf{H}^H \left(\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv} \right)^{-1}$$

Conclusions

- MMSE linear equalization is a well-studied field for combatting ISI channel.
- All MMSE equalizers share common feedforward filter structure:

- MMSE equalizer:
- Partial response and channel shortening equalizers:
- Decision feedback equalizer:

$$\mathbf{w}^H = \mathbf{e}_m^H \mathbf{R}_{xx} \mathbf{H}^H (\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv})^{-1}$$

$$\mathbf{w}^H = \tilde{\mathbf{b}}^H \mathbf{R}_{xx} \mathbf{H}^H (\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv})^{-1}$$

$$\mathbf{w}^H = (\sigma_x^2 \mathbf{e}_m^H + \mathbf{b}^H \mathbf{M} \mathbf{R}_{xx}) \mathbf{H}^H (\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv})^{-1}$$

- All filters first equalize the channel with $\mathbf{R}_{xx} \mathbf{H}^H (\mathbf{H} \mathbf{R}_{xx} \mathbf{H}^H + \mathbf{R}_{vv})^{-1}$, then reshape the IR with either $\tilde{\mathbf{b}}^H$ or $(\sigma_x^2 \mathbf{e}_m^H + \mathbf{b}^H \mathbf{M} \mathbf{R}_{xx})$. This decreases the ISI due to the extra degree of freedom provided by these IRs.
- There is an inherent whitening matched filter front end in an equalizer designed for the MMSE criterion:

$$\mathbf{w}^H = \mathbf{e}_m^H \mathbf{b}^H \left(\mathbf{R}_{xx}^{-1} + \mathbf{H}^H \mathbf{R}_{vv}^{-1} \mathbf{H} \right)^{-1} \underbrace{\mathbf{H}^H \mathbf{R}_{vv}^{-1}}_{\text{whitening matched filter}}$$

- Design of the feedback filter can be application specific.

References

- [1]: Simon Haykin, Adaptive Filter Theory, Prentice Hall, 2001,
- [2]: Al-Dhahir, “FIR Channel-Shortening Equalizers for MIMO ISI Channels”, IEEE Trans. Commun., vol. 49, Feb. 2001, pp. 213-8,
- [3]: R. A. Casas, et al, DFE Tutorial, <http://www.ece.osu.edu/~schniter/postscript/dfetutorial.pdf>,
- [4]: M. Brookes, Matrix Reference Manual, <http://www.ee.ic.ac.uk/hp/staff/dmb/matrix/intro.html>.