

Lecture 15 Decision Feedback Equalizers February 29, 2024

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Announcements & Agenda

- Announcements
 - PS7 due next Wed (March 6) last required
 - PS8 is optional (but relatively easy) and due March 12 (solutions immediate)

Today

- Continue Decision Feedback MMSE
- Examples
- The Whitened Matched Filter and Zero Forcing
- FIR Implementation



Decision Feedback (successive decoding)

Section 3.6

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Add previous-decision use



• MMSE optimizes both feedforward, W(D), and feedback, B(D), filters.

Definition 3.6.1 [Mean Square Error (for DFE)] The MMSE-DFE error signal is $e_k = x_k - z'_k$. (3.218)

The MMSE for the MMSE-DFE is

$$\sigma_{_{MMSE-DFE}}^2 \stackrel{\Delta}{=} \min_{w_k, b_k} \mathbb{E}\left[|x_k - z'_k|^2\right] \quad . \tag{3.219}$$

- Subtraction of ISI eliminates noise enhancement (almost entirely, even though only "past" ISI),
 - but lowers received message energy w.r.t. MFB.

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Sec 3.6.1

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MMSE-DFE Solution

• MMSE estimates are linear (see Appendix D), so for any given B(D) acting on x_k :

$$W_{MMSE-DFE}(D) = B(D) \cdot W_{MMSE-LE}(D) = \frac{B(D)}{\|h\| \cdot \left[Q(D) + \frac{1}{SNR_{MFB}}\right]}$$
Full math details in 3.6
$$E_{MMSE-DFE}(D) = B(D) \cdot E_{MMSE-LE}(D)$$
Further
$$R_{ee,dfe}(D) = B(D) \cdot R_{ee,le}(D) \cdot B^{*}(D^{-*})$$

$$E_{FE \ error \ autocorellation}$$

$$E_{E \ error \ autocorellation}$$

• Optimization considers all causal monic ($b_0 = 1$) B(D).

$$R_{ee,le}(D) = \frac{\frac{\mathcal{N}_0}{2}}{\|h\|^2 \cdot (Q(D) + 1/SNR_{^{MFB}})}$$



Sec 3.6.1

USE: Canonical (Spectral) Factorization (D.3.4)

Realizes a spectrum magnitude with white input into a causal, and 1-to-1 invertible, filter.

White
$$v_k$$
 $V(D)$ $G_{\chi}(D)$ $\mathcal{E}_{\chi} = \mathbb{E}[|\chi|^2]$
sequence x_k $X(D)$ $1/G_{\chi}(D)$ $V(D)$

• The power spectral density of x_k is $R_{xx}(D)$ with $D = e^{-j\omega T}$.

• Find $G_{\chi}(D)$?

$$R_{\chi\chi}(D) = \begin{array}{c} G_{\chi}(D) & \cdot S_{\chi} \cdot G_{\chi}^{*}(D^{-*}) \\ \overbrace{monic}^{monic} & \stackrel{\smile}{}_{so}^{>0} \\ min-phase \end{array}$$

 $G_x(D) = 1 + g_1 \cdot D + g_2 \cdot D^2 + \cdots$ All poles/zeros outside unit circle

 $G_x^*(D^{-*})=1+g_1^*\cdot D^{-1}+g_2^*\cdot D^{-2}+\cdots$ All poles/zeros inside unit circle

| Spectral f | actorization | |
|-------------------|--------------|---------|
| of R | $R_{xx}(D)$ | |
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Sec D.3.4

Paley Weiner Criterion

Well, but we cannot always do this factorization (even if it is a power spectral density); it must satisfy:

$$\frac{1}{2\pi} \int_{-\frac{\pi}{T}}^{\frac{\pi}{T}} |\ln R_{xx}(e^{-j\omega})| d\omega < \infty$$

No "dead zones"





Back to DFE MMSE (error autocorrelation)

$$Q(D) + \frac{1}{SNR_{MFB}} = \gamma_0 \cdot G(D) \cdot G^*(D^{-*})$$

• Always satisfies PWC for finite $SNR_{MFB} > 0$.

$$\bar{R}_{ee}(D) = \frac{B(D) \cdot B^*(D^{-*})}{Q(D) + \frac{1}{SNR_{MFB}}} \cdot \frac{\frac{N_0}{2}}{\|h\|^2}$$
$$= \frac{B(D)}{G(D)} \cdot \frac{B^*(D^{-*})}{G^*(D^{-*})} \cdot \frac{\frac{N_0}{2}}{\gamma_0 \cdot \|h\|^2} = \frac{\frac{N_0}{2} \cdot \|b/g\|^2}{\gamma_0 \cdot \|h\|^2}$$

The fractional polynomial inside the squared norm b/g is necessarily monic and causal, and therefore the squared norm has a minimum value of 1.

Minimum when B(D) = G(D)

Minimum value for MSE-DFE

 $\overline{r}_{ee,0} \geq \frac{\overline{2}}{\gamma_0 \cdot \|h\|^2}$

Sec 3.6.1

MMSE-DFE Best Settings

Detailed math to check previous slides sketch is in Section 3.6.

Lemma 3.6.1 [MMSE-DFE] The MMSE-DFE has feedforward section

$$W(D) = \frac{1}{\|h\| \cdot \gamma_0 \cdot G^*(D^{-*})}$$

(realized with delay, as it is strictly noncausal) and feedback section

$$B(D) = G(D) \tag{3.233}$$

where G(D) is the unique canonical factor of:

$$Q(D) + \frac{1}{SNR_{MFB}} = \gamma_0 \cdot G(D) \cdot G^*(D^{-*}) \quad . \tag{3.234}$$

This text also calls the joint matched-filter/sampler/W(D) combination in the forward path of the DFE the "Mean-Square Whitened Matched Filter (MS-WMF)". These settings for the MMSE-DFE minimize the MSE as was shown above. MS-WMF = WMF, if infinite SNR



(3.232)

The MMSE value itself

Take log of factorization:

$$\frac{T}{2\pi} \int_{-\frac{\pi}{T}}^{\frac{\pi}{T}} \ln\left(Q(e^{-j\omega T}) + \frac{1}{SNR_{MFB}}\right) \cdot d\omega = \ln(\gamma_0) + \frac{T}{2\pi} \int_{-\frac{\pi}{T}}^{\frac{\pi}{T}} \ln\left(G(e^{-j\omega T}) \cdot G^*(e^{-j\omega T})\right) \cdot d\omega$$
$$= \ln(\gamma_0) \quad . \tag{3.236}$$

Salz Formula (1977):

$$\sigma_{_{MMSE-DFE}}^2 = \frac{\frac{N_0}{2}}{\|h\|^2} \cdot e^{-\frac{T}{2\pi} \int_{-\frac{\pi}{T}}^{\frac{\pi}{T}} \ln\left(Q(e^{-j\omega T}) + \frac{1}{SNR_{MFB}}\right) \cdot d\omega}$$

Finally, back to our receiver SNR-based analysis:

$$SNR_{MMSE-DFE} = \frac{\bar{\mathcal{E}}_{\boldsymbol{x}}}{\sigma_{MMSE-DFE}^{2}} = \gamma_{0} \cdot SNR_{MFB}$$

$$SNR_{MMSE-DFE,U} = SNR_{MMSE-DFE} - 1$$

$$\leq SNR_{MFB}$$
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Unbiased MMSE-DFE

Final result



$$G_U(D) = 1 + \frac{SNR_{MMSE-DFE}}{SNR_{MMSE-DFE,U}} \cdot [G(D) - 1]$$

- Can absorb unbiasing multiply into feedforward filter
- Must always perform at least as well as LE, why?



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Sec 3.6.2 PS7.2 (NP-DFE)

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Examples

Section 3.6.4

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Return to $H(D) = 1 + .9 \cdot D^{-1}$

- Refresh $||h||^2 = 1.81$, $SNR_{MFB} = 10$ $H(\omega) = \begin{cases} \sqrt{T} \cdot (1 + .9e^{j\omega T}) & |\omega| \le \frac{\pi}{T} \\ 0 & |\omega| > \frac{\pi}{T} \end{cases}$ $Q(D) + \frac{1}{10} = \left(\frac{1}{1.81}\right) \cdot \left[9 \cdot D^{-1} + 1.991 + .9 \cdot D\right]$ $\frac{1}{||h||^2}$ % Q is positive real – roots are in conjugate-reciprocal pairs. >> roots([.9 1.991 .9]) = -1.5788 $=.785 \cdot (1 + .6334 \cdot D) \cdot (1 + .6334 \cdot D^{-1})$ -0.6334 γ_0 G(D)=B(D) $G(D^{-1})$ >> .9/(1.81*.6334) = 0.7850 % follows from last (or first) coefficient $W(D) = \frac{1}{\sqrt{1.81}(.785)(1 + .633 \cdot D^{-1})} = \frac{.9469}{1 + .633 \cdot D^{-1}}$
 - Performance $SNR_{MMSE-DFE,U} = .785 \cdot 10 1 = 6.85 \cdot (8.4 \, dB)$
 - 1.6 dB less than MFB –
 - $G_U(D) = (1 + .7259 \cdot D)$



Feb 29, 2024 PS7.1 (DFE better) Sec 3.6.4

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PS7.5 (IIR)

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>> 7.85/6.85*.6334 = 0.7259 >> .9469*(7.85/6.85)= 1.0851

Example continued

FF filter magnitude Feedback filter magnitude 10 5 magnitude in dB magnitude in dB -5 -0.5 -0.4 -0.3 -0.1 0.1 0.2 0.3 0.4 0.5 -0.2 0 -10 -0.5 -0.4 -0.3 -0.2 -0.1 0.2 0.3 0.4 0 0.1 0.5 normalized frequency normalized frequency

- FF is single pole (anticausal) IIR filter, while FB is two-tap causal FIR filter (really 1 tap implemented).
- The anticausal FF is implemented by FIR approximation with delay.



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Sec 3.6.4

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Eliminates Noise Enhancement

MS-WMF filter magnitude 0 -5 magnitude in dB -10 -15 -0.5 -0.3 -0.2 0.2 -0.4 -0.1 0 0.1 0.3 0.4 0.5 normalized frequency

- The MS-WMF is nearly flat (so noise/error stays pretty flat, and error itself is white.
- It is nearly an "all-pass" filter (MS-WMF just adjusts phase to minimum phase channel):
 - Minimum phase channel has maximum energy at left for all phase equivalents (helpful for best "feedback" to have no advance ISI).



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Sec 3.6.4

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Includes MF also

Add a Code?

- BICM, 64-state code with 4PAM
 - Maintains $\overline{b} = 1$, $r = \frac{1}{2}$
- $\mathcal{E}_{4PAM} = 5 = 5 \cdot \mathcal{E}_{2PAM}$, so 7dB loss
- *d*_{free} = 10, and each code gets *r* =½ resources
 - $r \cdot d_{free} \cdot SNR = 6.4 + 7 = 13.4 \text{ dB}$

+3 • 10
+1 • 11
$$d = 4$$

 $-1 • 01 d = 4$
 $-3 • 00$

Thanks to DFE we've maintained the data rate And reduced error prob to $P_e \cong 10^{-6}$

- See a problem perhaps?
 - Decision delay of code means FB input could be incorrect









 $G(D) = (1 - (.4502 - j.0276) \cdot D) \cdot (1 - (-.0276 + .4502j) \cdot D)$

• $SNR_{MMSE-DFE,U} = .7865 \cdot 10 - 1 = 7.88 \cdot (\sim 8.4 \text{ dB} \text{ also} - \text{coincidence}).$





Sec 3.6.4

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Whitened Matched Filter and the ZF-DFE

Section 3.6.3

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ZF-DFE

- Set $SNR_{MFB} = \infty$, so then the spectral factorization is directly of $Q(D) = \eta_0 \cdot P_c(D) \cdot P_c^*(D^{-*})$.
 - $B(D) = P_c(D)$; $\eta_0 = \frac{1}{\|P_c\|^2}$
 - $W(D) = [\eta_0 \cdot ||h|| \cdot P_c^*(D^{-*})]^{-1}$
 - $SNR_{ZF-DFE} = \eta_0 \cdot SNR_{MFB}$
- However, the spectral factorization is not guaranteed to exist (might not satisfy PWC).
- Subtle mistake is "ZF-DFE has no noise enhancement so same as MMSE-DFE" ?
 - Zero noise enhancement is true ONLY if the entire transmit band is energized.
 - And still not as good as MMSE-DFE even then.

The issue is that best transmit spectrum almost never satisfies PWC

- We'll see as 379's progress that this optimized-input modulator needs care with DFEs.
 - Mistakes dwarf coding-gain improvements.
 - Indeed, using nonero-gap code magnifies the loss.



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Sec 3.6.3

PS7.3 (EPR) L15: 19

 $\overline{2T}$

Return to $H(D) = 1 + .9 \cdot D^{-1}$

• Refresh $||h||^2 = 1.81$, $SNR_{MFB} = 10$

$$H(\omega) = \begin{cases} \sqrt{T} \cdot \left(1 + .9e^{j\omega T}\right) & |\omega| \le \frac{\pi}{T} \\ 0 & |\omega| > \frac{\pi}{T} \end{cases}$$

$$Q(D) + \frac{1}{10} = \left(\frac{1}{1.81}\right) \cdot \left[9 \cdot D^{-1} + 1.81 + .9 \cdot D\right]$$
$$= .5525 \cdot (1 + .9 \cdot D) \cdot (1 + .9 \cdot D^{-1})$$

 $P_{c}(D) = B(D)$

$$W(D) = \frac{\sqrt{1.81}}{1 + .9 \cdot D^{-1}}$$
 $MWF = \frac{1 + .9 \cdot D}{1 + .9 \cdot D^{-1}}$ (all pass filter, exactly)

 $P_{C}(D^{-1})$

• Performance $SNR_{ZF-DFE} = .5525 \cdot 10 = 5.25 \cdot (7.4 \, dB)$

 η_0



Sec 3.6.3

L phases, FSE – same, but for FF only

There are l phases of input samples for each symbol-sample-time output ("polyphase system")

$$y_i(kT) = y(kT - iT/l)$$
, $i = 0, ..., l - 1$

- Each phase has a D-Transform $Y_i(D) = H_i(D) * X(D) + N_i(D)$
 - This is a form of what is called "diversity" where several channels carry the same input to different outputs

$$\underline{\mathbf{Y}(D)} = \underline{\mathbf{H}(D)} \cdot X(D) + \underline{\mathbf{N}(D)}$$
$$\underbrace{\mathbf{W}(D)}_{l \times 1} \quad \underbrace{\mathbf{W}(D)}_{l \times 1}$$

• Retains all timing-offset benefits, matched-filter absorption, etc.



FIR Implementation

Section 3.7

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Sample Fast enough – absorb MF



- Continue polyphase channel model ^l/_T where l > 1:
 - Initial example is integer (more generally rational fraction).
 - Channel model is FIR (like with DMT earlier).

$$y(k - iT/l) = \sum_{m=k-\nu}^{k} x_m \cdot h\left(kT - \frac{[l-1]T}{l} - mT\right) + n(k - iT/l)$$

> 1:
on).

$$x_k$$

 $n(k - iT/l)$
 $h(kT - [l - 1]T/l)$
 $h(kT - [l - 1]T/l)$

Each mini channel corresponds to one of the *l* sampling phases per symbol period.



Sec 3.7.1

L16: 23

Creating a matrix FIR channel

• Creates a vector channel
$$y_k = \sum_{m=k-\nu}^k x_m \cdot h_{k-m} + n_k = \sum_{m=0}^{\nu} x_{k-m} \cdot h_m + n_k$$

• where $y_k = \begin{bmatrix} y(kT) \\ y(k-T/l) \\ \vdots \\ y(k-(l+1)T/l) \end{bmatrix} \quad h_k = \begin{bmatrix} h(kT) \\ h(k-T/l) \\ \vdots \\ h(k-(l+1)T/l) \end{bmatrix} \quad n_k = \begin{bmatrix} n(kT) \\ n(k-T/l) \\ \vdots \\ n(k-(l+1)T/l) \end{bmatrix}$
• channel model $y_k = \begin{bmatrix} h_0 h_1 \cdots h_{\nu} \end{bmatrix} \cdot \begin{bmatrix} x_k \\ x_{k-1} \\ \vdots \\ x_{k-\nu} \end{bmatrix} + n_k$
• the values in the FIR equalizer's span are x_k
 $Y_k = \begin{bmatrix} y_k \\ y_{k-1} \\ \vdots \\ y_{k-N_f+1} \end{bmatrix} = \begin{bmatrix} h_0 h_1 \cdots h_{\nu} & 0 \cdots & 0 \\ 0 & h_0 h_1 \cdots h_{\nu} & 0 & 0 \\ 0 & \cdots & 0 & h_0 h_1 & \cdots & h_{\nu} \end{bmatrix} \cdot X_k + N_k$
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Delay Δ – FIR is causal, estimates \hat{x}_{k-A}

Implemented receivers must have delay Δ (making FIR filters causal)



• Try to estimate $x_{k-\Delta}$

Not (usually) possible to have $\triangle = 0$

Precise delay may need to be optimized by trying a few (or via algorithm later).



Sec 3.7.1

L15:25

MMSE-LE Design

- The error signal is $e_k = x_{k-\Delta} z_k$; $\sigma^2_{MMSE-LE} = \mathbb{E}[|e_k|^2]$
- Orthogonality principle: $\mathbb{E}[e_k \cdot Y_k] = 0 \rightarrow MMSE$
 - $w = R_{xY} \cdot R_{YY}^{-1}$ and
 - $R_{xY} = \mathbb{E}[x_{k-\Delta} \cdot Y_k^*]$ • $R_{YY} = \mathbb{E}[Y_k \cdot Y_k^*]$

•
$$R_{xY} = \overline{\mathcal{E}}_x \cdot H^*_{\Delta+1}$$
 basically energy $x (\Delta + 1)^{th}$ row of H^*
• $R_{YY} = \overline{\mathcal{E}}_x \cdot H \cdot H^* + l \cdot \frac{N_0}{2} \cdot R_{NN}$

$$\boldsymbol{w} = H_{\Delta+1}^* \cdot \left(\boldsymbol{H} \cdot \boldsymbol{H}^* + \frac{l}{SNR} \cdot \boldsymbol{R}_{NN} \right)^{-1}$$

$$\sigma_{MMSE-LE}^2 = \bar{\mathcal{E}}_x \cdot \boldsymbol{w} \cdot \boldsymbol{R}_{\boldsymbol{Y}x}$$



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Sec 3.7.1

Detailed algebra in 3.7.1

 $SNR_{MMSE-LE} = \frac{\bar{\mathcal{E}}_{x}}{\sigma_{MMSE-LE}^{2}}$ $SNR_{MMSE-LE,U} = SNR_{MMSE-LE} - 1$ $\gamma_{MMSE-LE} = \frac{SNR_{MMSE-LE,U}}{SNR_{MFB}}$

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Return to 1+.9D⁻¹ example

• Oversampling l = 1, 3 taps $N_f = 3$, $\overline{\mathcal{E}}_x = 1$, and $\nu = 1$



$\Delta = \text{best value (by trial & error)}$



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Sec 3.7.3

3-tap FIR MMSE-LE and output



This FIR MMSE LE tries to flatten channel, but not enough taps.



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7-taps looks better



There will be diminishing performance improvement as length increases.



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Sec 3.7.3

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For the complex example



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Sec 3.7.3

FIR ZFE

Design applies MMSE-LE with infinite SNR.

• Finite taps cannot guarantee zero ISI, so there is:

$$\sigma_{ZFE-ISI}^2 = \bar{\mathcal{E}}_x - \boldsymbol{w} \cdot \boldsymbol{R}_{\boldsymbol{Y}x}$$

>> wzfe=H(:,4)'*inv(H*H') = 0.2702 -0.5434 0.8227

>> MMSEzf=1-wzfe*H(:,4) = 0.1773

• The FIR ZFE can still be biased – look at position of $x_{k-\Delta}$ so $w \cdot H$ in position $\Delta + 1$

chan=wzfe*H = 0.2432 -0.2189 0.1970 0.8227

- Bias removal thus inverts, so $\omega_{FIR-ZFE} = 1 / \mathbf{w} \cdot \mathbf{H} \cdot \mathbf{1}_{\Delta}$ this means residual ISI and noise increase by this factor also
- Analysis must also add the scale the enhanced noise $\sigma_{ZFE}^2 = \mathbf{w} \cdot R_{\mathbf{nn}} \cdot \mathbf{w}^*$
- Total is $\sigma_{FIR-ZFE}^2 = \omega_{FIR-ZFE} \cdot (\bar{\mathcal{E}}_x \mathbf{w} \cdot R_{\mathbf{Y}x} + \mathbf{w} \cdot R_{\mathbf{nn}} \cdot \mathbf{w}^*)$





End Lecture 15