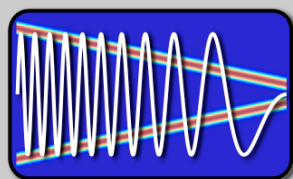


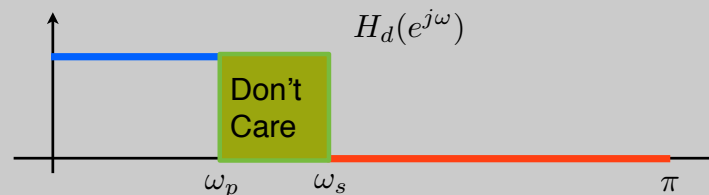
EE123



Digital Signal Processing

Lecture 23

Optimality



- Least Squares:

$$\text{minimize } \int_{\omega \in \text{care}} |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

Variation: weighted least-squares

$$\text{minimize } \int_{-\pi}^{\pi} W(\omega) |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

Design Through Optimization

- Idea: Sample/discretize the frequency response

$$H(e^{j\omega}) \Rightarrow H(e^{j\omega_k})$$

– Sample points are fixed $\omega_k = k \frac{\pi}{P}$

$$-\pi \leq \omega_1 < \dots < \omega_p \leq \pi$$

- M+1 is the filter order
- $P \gg M + 1$ (rule of thumb $P=15M$)
- Yields a (good) approximation of the original problem

Example: Least Squares

- Target: Design $M+1 = 2N+1$ filter
- First design non-causal $\tilde{H}(e^{j\omega})$ and hence $\tilde{h}[n]$
- Then, shift to make causal

$$h[n] = \tilde{h}[n - M/2]$$

$$H(e^{j\omega}) = e^{-j\frac{M}{2}} \tilde{H}(e^{j\omega})$$

Example: Least Squares

- Matrix formulation:

$$\tilde{h} = [\tilde{h}[-N], \tilde{h}[-N+1], \dots, \tilde{h}[N]]^T$$

$$b = [H_d(e^{j\omega_1}), \dots, H_d(e^{j\omega_P})]^T$$

$$A = \begin{bmatrix} e^{-j\omega_1(-N)} & \dots & e^{-j\omega_1(+N)} \\ \vdots & & \\ e^{-j\omega_P(-N)} & \dots & e^{-j\omega_P(+N)} \end{bmatrix}$$

$$\operatorname{argmin}_{\tilde{h}} \|A\tilde{h} - b\|^2$$

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Least Squares

$$\operatorname{argmin}_{\tilde{h}} \|A\tilde{h} - b\|^2$$

Solution:

$$\tilde{h} = (A^* A)^{-1} A^* b$$

- Result will generally be non-symmetric and complex valued.
- However, if $\tilde{H}(e^{j\omega})$ is real, $\tilde{h}[n]$ should have symmetry!

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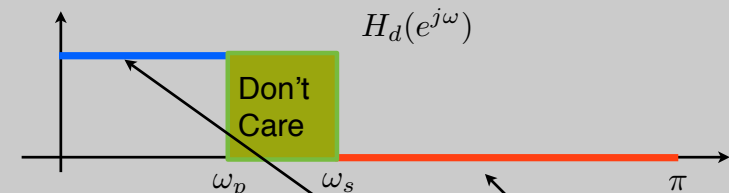
Design of Linear-Phase L.P Filter

- Suppose:
 - $\tilde{H}(e^{j\omega})$ is real-symmetric
 - M is even (M+1 taps)
- Then:
 - $\tilde{h}[n]$ is real-symmetric around midpoint
- So:

$$\begin{aligned} \tilde{H}(e^{j\omega}) &= \tilde{h}[0] + \tilde{h}[1]e^{-j\omega} + \tilde{h}[-1]e^{+j\omega} \\ &\quad + \tilde{h}[2]e^{-j2\omega} + \tilde{h}[-2]e^{+j2\omega} \dots \\ &= \tilde{h}[0] + 2 \cos(\omega)\tilde{h}[1] + 2 \cos(2\omega)\tilde{h}[2] + \dots \end{aligned}$$

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Least-Squares Linear-Phase Filter



Given M, ω_p , ω_s find the best LS filter:

$$A = \begin{bmatrix} 1 & \dots & 2 \cos(\frac{M}{2}\omega_1) \\ \vdots & & \\ 1 & \dots & 2 \cos(\frac{M}{2}\omega_p) \\ 1 & \dots & 2 \cos(\frac{M}{2}\omega_s) \\ \vdots & & \\ 1 & \dots & 2 \cos(\frac{M}{2}\omega_P) \end{bmatrix}$$

$$b = [1, 1, \dots, 1, 0, 0, \dots, 0]^T$$

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Least-Squares Linear-Phase Filter

Given M , ω_p , ω_s find the best LS filter:

$$A = \begin{bmatrix} 1 & \cdots & 2 \cos(\frac{M}{2}\omega_1) \\ \vdots & & \\ 1 & \cdots & 2 \cos(\frac{M}{2}\omega_p) \\ 1 & \cdots & 2 \cos(\frac{M}{2}\omega_s) \\ \vdots & & \\ 1 & \cdots & 2 \cos(\frac{M}{2}\omega_P) \end{bmatrix} \quad b = [1, 1, \dots, 1, 0, 0, \dots, 0]^T$$

$$\tilde{h}_+ = [\tilde{h}[0], \dots, \tilde{h}[\frac{M}{2}]]^T = (A^* A)^{-1} A^* b$$

$$\tilde{h} = \begin{cases} \tilde{h}_+[n] & n \geq 0 \\ \tilde{h}_+[-n] & n < 0 \end{cases}$$

$$h[n] = \tilde{h}[n - M/2]$$

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Extension:

- LS has no preference for pass band or stop band
- Use weighting of LS to change ratio

want to solve the discrete version of:

$$\text{minimize} \int_{-\pi}^{\pi} W(\omega) |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

where $W(\omega)$ is δ_p in the pass band and δ_s in stop band

Similarly: $W(\omega)$ is 1 in the pass band and δ_p/δ_s in stop band

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Weighted Least-Squares

$$\text{argmin}_{\tilde{h}_+} (A\tilde{h}_+ - b)^* W^2 (A\tilde{h}_+ - b)$$

Solution:

$$\tilde{h}_+ = (A^* W^2 A)^{-1} W^2 A^* b$$

$$W = \begin{bmatrix} 1 & & & & & & 0 \\ & 1 & & & & & \\ & & \cdots & & & & \\ & & & \frac{\delta_p}{\delta_s} & & & \\ & & & & \cdots & & \\ 0 & & & & & & \frac{\delta_p}{\delta_s} \end{bmatrix}$$

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Min-Max optimal Filters

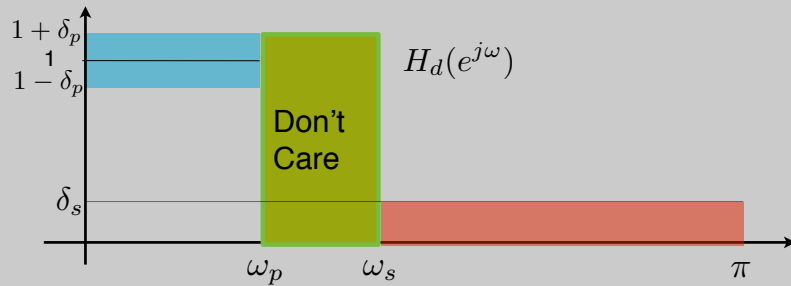
- Chebychev Design (min-max)

$$\text{minimize}_{\omega \in \text{care}} \max |H(e^{j\omega}) - H_d(e^{j\omega})|$$

- Parks-McClellan algorithm - equi-ripple
- Also known as Remez exchange algorithms (signal.remez)
- Also with convex optimization

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Specifications



- Filter specifications are given in terms of boundaries

Min-Max Filter Design

- Minimize:

- max pass-band ripple

$$1 - \delta_p \leq |H(e^{j\omega})| \leq 1 + \delta_p, \quad 0 \leq \omega \leq \omega_p$$

- min-max stop-band ripple

$$|H(e^{j\omega})| \leq \delta_s, \quad \omega_s \leq \omega \leq \pi$$

Min-max Ripple Design

- Recall, $\tilde{H}(e^{j\omega})$ is symmetric and real

- Given ω_p ω_s M , find δ, \tilde{h}_+ :

$$\text{minimize } \delta$$

Subject to :

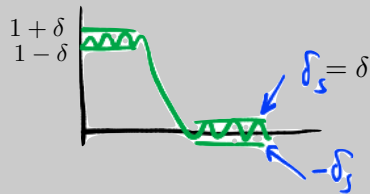
$$1 - \delta \leq \tilde{H}(e^{j\omega_k}) \leq 1 + \delta \quad 0 \leq \omega_k \leq \omega_p$$

$$-\delta \leq \tilde{H}(e^{j\omega_k}) \leq \delta \quad \omega_s \leq \omega_k \leq \pi$$

$$\delta > 0$$

- Solution is a linear program in δ, \tilde{h}_+

- A well studied class of problems



Min-Max Ripple via Linear Programming

minimize δ

subject to :

$$1 - \delta \preceq A_p \tilde{h}_+ \preceq 1 + \delta$$

$$-\delta \preceq A_s \tilde{h}_+ \preceq \delta$$

$$\delta > 0$$

$$A_p = \begin{bmatrix} 1 & 2 \cos(\omega_1) & \cdots & 2 \cos(\frac{M}{2}\omega_1) \\ \vdots & & & \\ 1 & 2 \cos(\omega_p) & \cdots & 2 \cos(\frac{M}{2}\omega_p) \end{bmatrix}$$

$$A_s = \begin{bmatrix} 1 & 2 \cos(\omega_s) & \cdots & 2 \cos(\frac{M}{2}\omega_s) \\ \vdots & & & \\ 1 & 2 \cos(\omega_p) & \cdots & 2 \cos(\frac{M}{2}\omega_p) \end{bmatrix}$$

capital P

Convex Optimization

- Many tools and Solvers
- Tools:
 - CVX (Matlab) <http://cvxr.com/cvx/>
 - CVXOPT, CVXMOD (Python)
- Engines:
 - Sedumi (Free)
 - MOSEK (commercial)
- Take EE127!

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Using CVX (in Matlab)

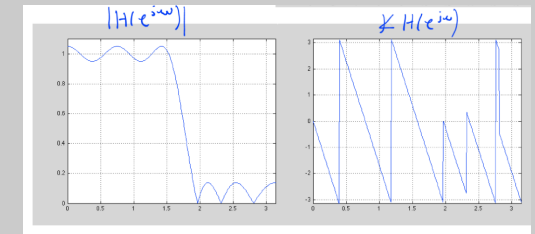
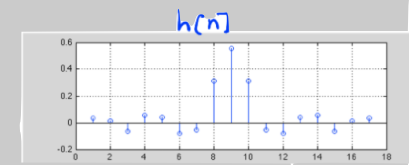
```
M = 16;  
wp = 0.5*pi;  
ws = 0.6*pi;  
MM = 15*M;  
w = linspace(0,pi,MM);
```

```
idxp = find(w <= wp);  
idxs = find(w >= ws);
```

```
Ap = [ones(length(idxp),1) 2*cos(kron(w(idxp)',  
[1:M/2]))];  
As = [ones(length(idxs),1) 2*cos(kron(w(idxs)',  
[1:M/2]))];
```

```
% optimization  
cvx_begin  
variable hh(M/2+1,1);  
variable d(1,1);
```

```
minimize(d)  
subject to  
Ap*hh <= 1+d;  
Ap*hh >= 1-d;  
As*hh < d;  
As*hh > -d;  
ds > 0;  
cvx_end  
h = [hh(end:-1:1); hh(2:end)];
```



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Variations:

- Convex Problems:
 - Fix δ_s optimize for δ_p
 - Fix δ_p optimize for δ_s
 - Linear constraints on $h[n]$
- Quasi-Convex (feasible through bisection)
 - Fix δ_p, δ_s, M , minimize $\Delta\omega = \omega_s - \omega_p$
 - Fix $\delta_p, \delta_s, \Delta\omega = \omega_s - \omega_p$, minimize M

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Bisection Example: Minimize M

- given $\delta_p, \delta_s, \Delta\omega = \omega_s - \omega_p$ Initialize problem with:
 - Set M_{\min} to be small and hence infeasible
 - Set M_{\max} to be large and hence feasible
 - Set $M = \text{floor}(M_{\max}/2 + M_{\min}/2)$
- Given $M, \delta_p, \Delta\omega = \omega_s - \omega_p$ solve for minimum δ_s
 - If δ_s violates constraints, set $M_{\min} = M$
 - if δ_s within constraints, set $M_{\max} = M$
 - Set $M = \text{floor}(M_{\max}/2 + M_{\min}/2)$
 - Repeat till M is tight

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IIR Design

- Historically
 - Continuous IIR design was advanced
 - Use results from C.T to D.T
 - C.T IIT designs have closed form, easy to use
 - Easy to control Magnitude, not easy to control phase
- Common Types:
 - Butterworth - monotonic, no ripple
 - Chebyshev - Type I, pass band ripple, Type II stop band ripple
 - Elliptic - Ripples in both bands

Design of D.T IIR Filters from Analog

- Discretize by one of many techniques
- $H_c(s) \Rightarrow H(z)$
- Must satisfy:
 - Imaginary axis is mapped to unit circle
 - stability of $H_c(s)$ should result in stable $H(z)$
- Two methods:
 - Impulse invariance - match impulse response
 - Bilinear transformation