

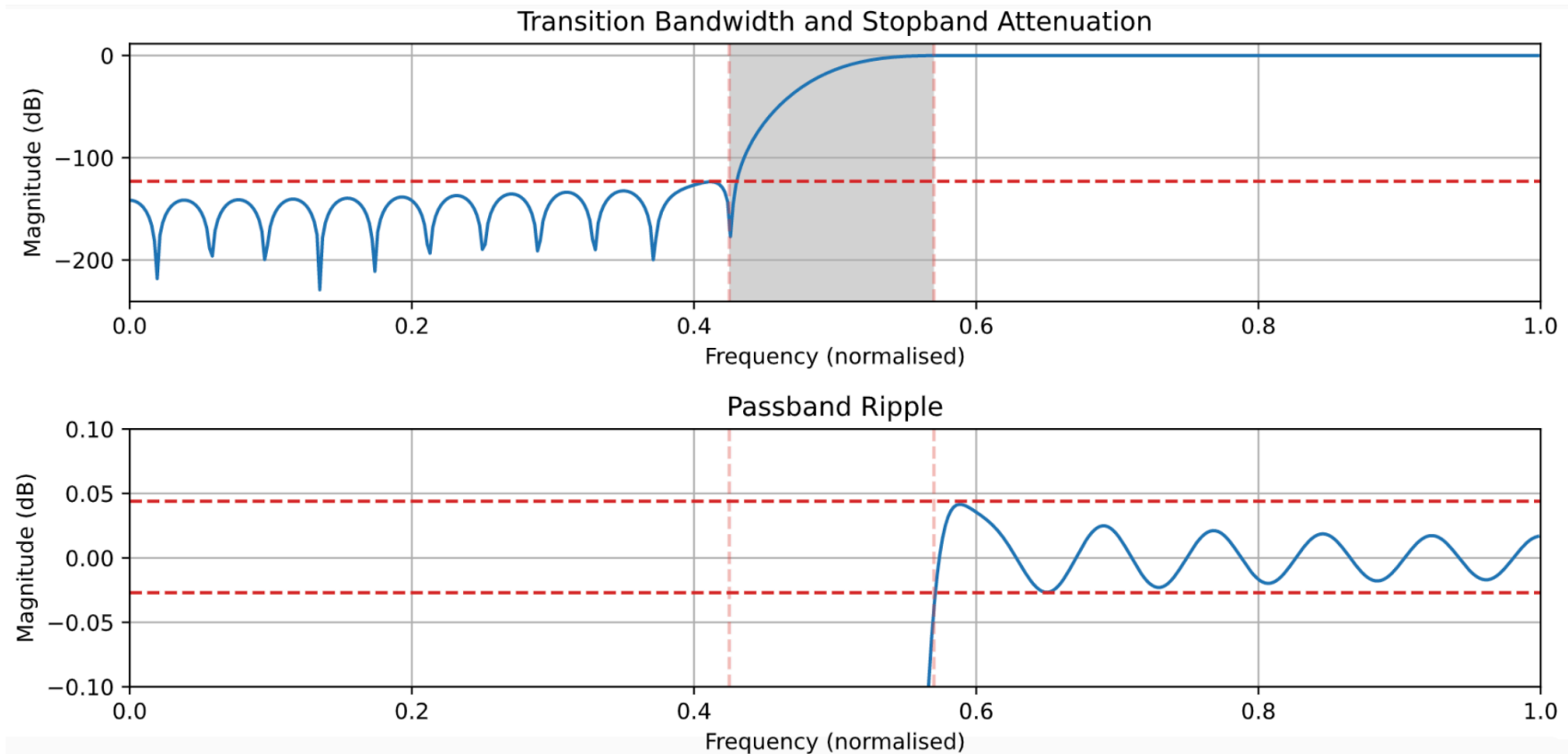
# ESE 5310: Digital Signal Processing

---

Lecture 17: March 23, 2024  
Optimal Filter Design



# Window Method Demo





# Optimal Filter Design

---

- ❑ Window method
  - Design Filters heuristically using windowed sinc functions
  - Choose order and window type
  - Check DTFT to see if filter specs are met
- ❑ Optimal design
  - Design a filter  $h[n]$  with  $H(e^{j\omega})$
  - Approximate  $H_d(e^{j\omega})$  with some optimality criteria - or satisfies specs (ie. Constraints and objective)



# Mathematical Optimization

---

## (mathematical) optimization problem

$$\begin{array}{ll}\text{minimize} & f_0(x) \\ \text{subject to} & f_i(x) \leq b_i, \quad i = 1, \dots, m\end{array}$$

- $x = (x_1, \dots, x_n)$ : optimization variables
- $f_0 : \mathbf{R}^n \rightarrow \mathbf{R}$ : objective function
- $f_i : \mathbf{R}^n \rightarrow \mathbf{R}, i = 1, \dots, m$ : constraint functions

**optimal solution**  $x^*$  has smallest value of  $f_0$  among all vectors that satisfy the constraints



# Solving Optimization Problems

---

## **general optimization problem**

- very difficult to solve
- methods involve some compromise, *e.g.*, very long computation time, or not always finding the solution

**exceptions:** certain problem classes can be solved efficiently and reliably

- least-squares problems
- linear programming problems
- convex optimization problems



# Least-Squares Optimization

---

$$\text{minimize } \|Ax - b\|_2^2$$

## **solving least-squares problems**

- analytical solution:  $x^* = (A^T A)^{-1} A^T b$
- reliable and efficient algorithms and software
- computation time proportional to  $n^2 k$  ( $A \in \mathbf{R}^{k \times n}$ ); less if structured
- a mature technology

## **using least-squares**

- least-squares problems are easy to recognize
- a few standard techniques increase flexibility (*e.g.*, including weights, adding regularization terms)



# Linear Programming

---

$$\begin{array}{ll}\text{minimize} & c^T x \\ \text{subject to} & a_i^T x \leq b_i, \quad i = 1, \dots, m\end{array}$$

## **solving linear programs**

- no analytical formula for solution
- reliable and efficient algorithms and software
- computation time proportional to  $n^2m$  if  $m \geq n$ ; less with structure
- a mature technology

## **using linear programming**

- not as easy to recognize as least-squares problems
- a few standard tricks used to convert problems into linear programs  
(*e.g.*, problems involving  $\ell_1$ - or  $\ell_\infty$ -norms, piecewise-linear functions)



# Convex Optimization

---

$$\begin{array}{ll}\text{minimize} & f_0(x) \\ \text{subject to} & f_i(x) \leq b_i, \quad i = 1, \dots, m\end{array}$$

- objective and constraint functions are convex:

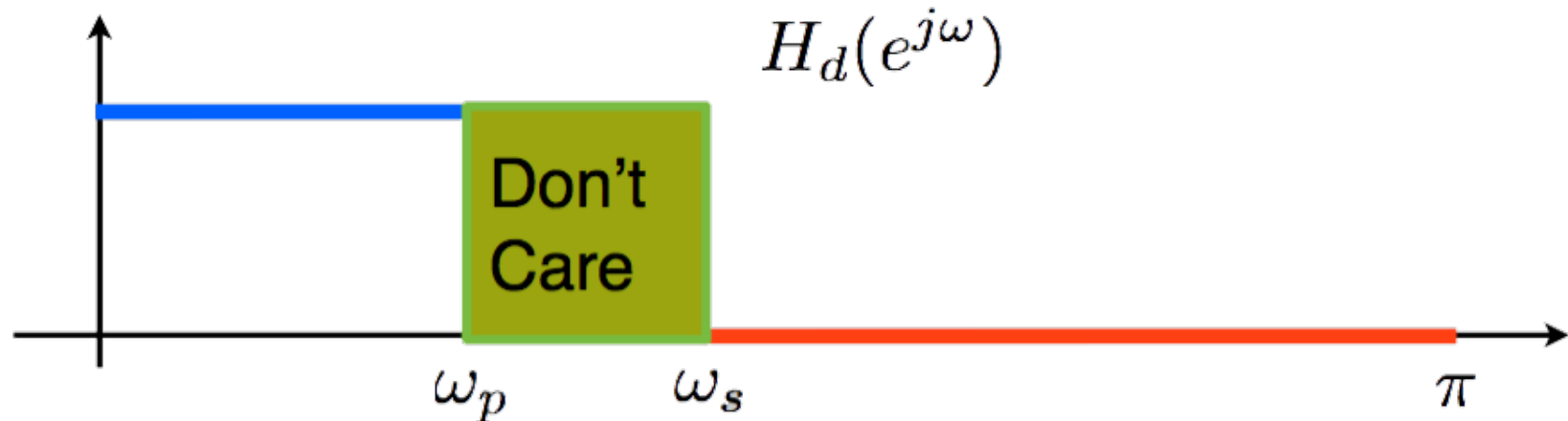
$$f_i(\alpha x + \beta y) \leq \alpha f_i(x) + \beta f_i(y)$$

if  $\alpha + \beta = 1$ ,  $\alpha \geq 0$ ,  $\beta \geq 0$

- includes least-squares problems and linear programs as special cases



# Optimality – Least Squares



□ 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  $\tilde{h}[n]$



# 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

---

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

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

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

$$\operatorname{argmin}_{\tilde{h}} ||A\tilde{h} - b||_2^2$$



# Least-Squares

---

$$\operatorname{argmin}_{\tilde{h}} \quad ||A\tilde{h} - b||_2^2$$

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!

# Design of Linear-Phase L.P Filter

□ Suppose:

- $\tilde{H}(e^{j\omega})$  is real-symmetric
- $M$  is even ( $M+1$  length)

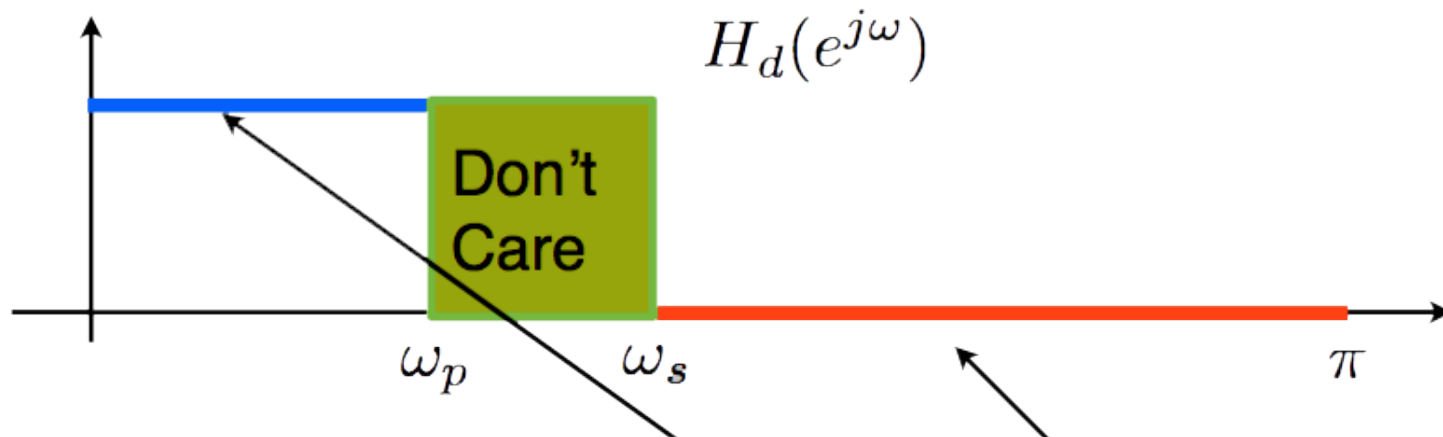
□ Then:

- $\tilde{h}[n]$  is real-symmetric around midpoint

□ So:

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

# Least-Squares Linear Phase Filter



Given  $M$ ,  $\omega_P$ ,  $\omega_s$  find the best LS filter:

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

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

Capital P



# Least-Squares Linear Phase Filter

Given  $M$ ,  $\omega_P$ ,  $\omega_s$  find the best LS filter:

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

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

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

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



## 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  $\delta_p$  in the pass band and  $\delta_s$  in stop band

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



# Weighted Least-Squares

---

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

Solution:

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

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



# Optimality – min-max

---

## □ Chebychev Design (min-max)

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

- Parks-McClellan algorithm - equiripple
- Also known as Remez exchange algorithms (signal.remez)
- Can also use convex optimization

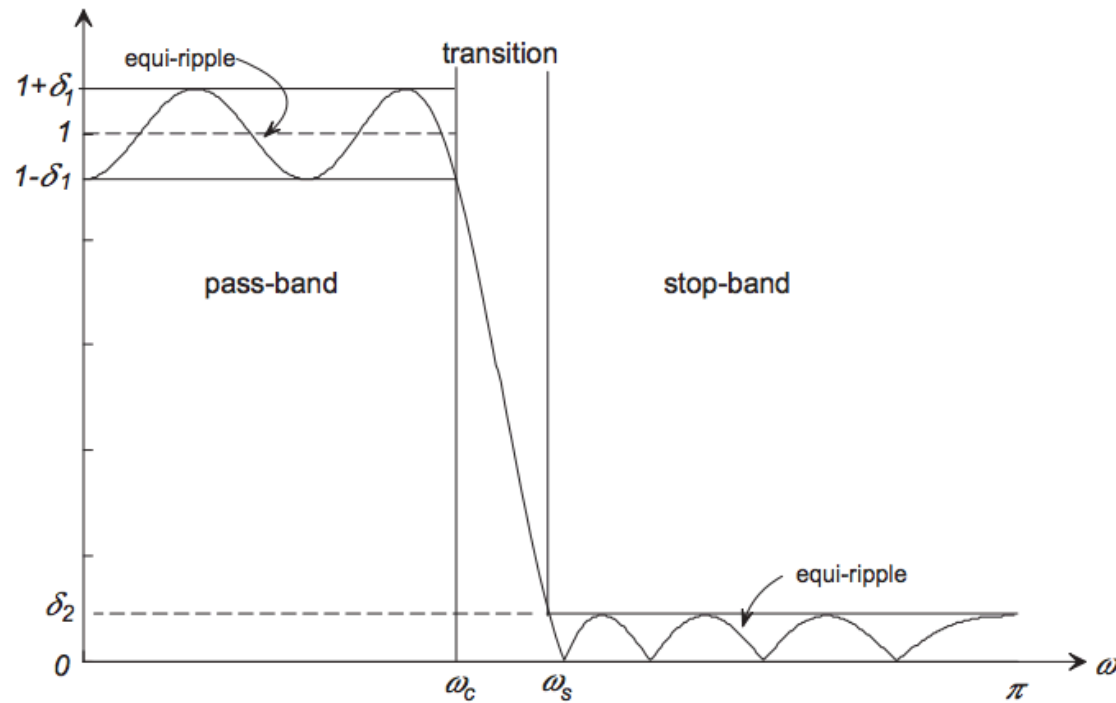


# Parks-McClellan

---

- ❑ Allows for multiple pass- and stop-bands.
- ❑ Is an equi-ripple design in the pass- and stop-bands, but allows independent weighting of the ripple in each band.
- ❑ Allows specification of the band edges.

# Parks-McClellan: LP Filter



□ For the low-pass filter shown above the specifications are

$$\begin{aligned} 1 - \delta_1 &< H(e^{j\omega}) < 1 + \delta_1 && \text{in the pass-band } 0 < \omega \leq \omega_c \\ -\delta_2 &< H(e^{j\omega}) < \delta_2 && \text{in the stop-band } \omega_s < \omega \leq \pi. \end{aligned}$$



# Parks-McClellan: LP Filter

---

- ❑ Need to determine  $M+1$  (length of the filter) and the filter coefficients  $\{h_n\}$



# Parks-McClellan: LP Filter

---

- ❑ Need to determine  $M+1$  (length of the filter) and the filter coefficients  $\{h_n\}$
- ❑ If we assume  $M$  even and even symmetry FIR filter (Type I), then

$$A_e(e^{j\omega}) = h_e[0] + \sum_{n=1}^L 2h_e[n] \cos(\omega n).$$

$$H(e^{j\omega}) = A_e(e^{j\omega})e^{-j\omega M/2}.$$





# Parks-McClellan: LP Filter

---

□ Reformulate:

$$A_e(e^{j\omega}) = h_e[0] + \sum_{n=1}^L 2h_e[n] \cos(\omega n).$$

□ ...to fitting a polynomial

$$A_e(e^{j\omega}) = \sum_{k=0}^L a_k (\cos \omega)^k,$$

$$A_e(e^{j\omega}) = P(x)|_{x=\cos \omega}, \quad P(x) = \sum_{k=0}^L a_k x^k.$$



# Parks-McClellan: LP Filter

---

- Define approximation error function

$$E(\omega) = W(\omega)[H_d(e^{j\omega}) - A_e(e^{j\omega})],$$

$$W(e^{j\omega}) = \begin{cases} \delta_2/\delta_1 & \text{in the pass-band} \\ 1 & \text{in the stop-band} \\ 0 & \text{in the transition band.} \end{cases}$$



# Parks-McClellan: LP Filter

---

- Define approximation error function

$$E(\omega) = W(\omega)[H_d(e^{j\omega}) - A_e(e^{j\omega})],$$

- Apply min-max or Chebyshev criteria

$$\min_{\{h_e[n]: 0 \leq n \leq L\}} \left( \max_{\omega \in F} |E(\omega)| \right),$$

# Parks-McClellan – Alternation Theorem

- The algorithm uses Chebyshev's alternation theorem to recognize the optimal solution.

Define the error  $E(x)$  as above, namely

$$E(e^{j\omega}) = W(e^{j\omega}) (H_d(e^{j\omega}) - H(e^{j\omega}))$$

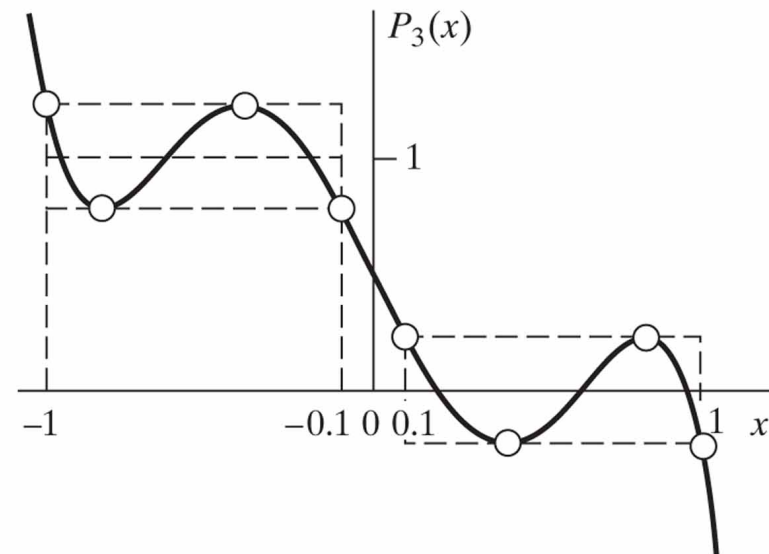
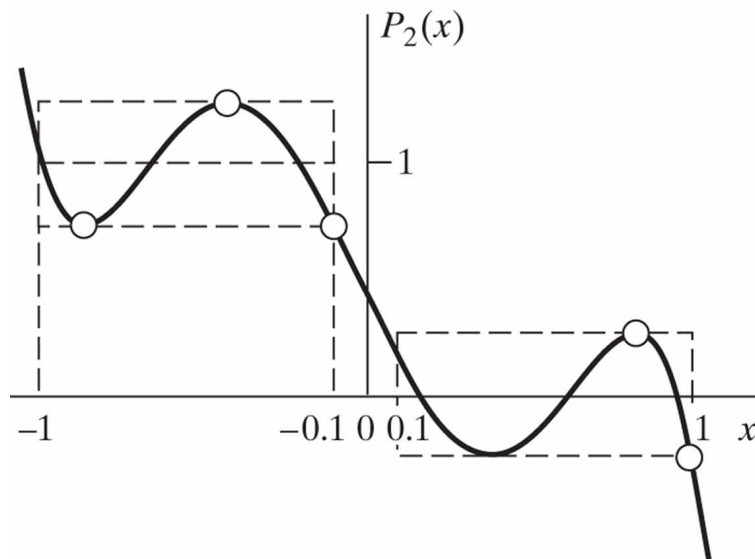
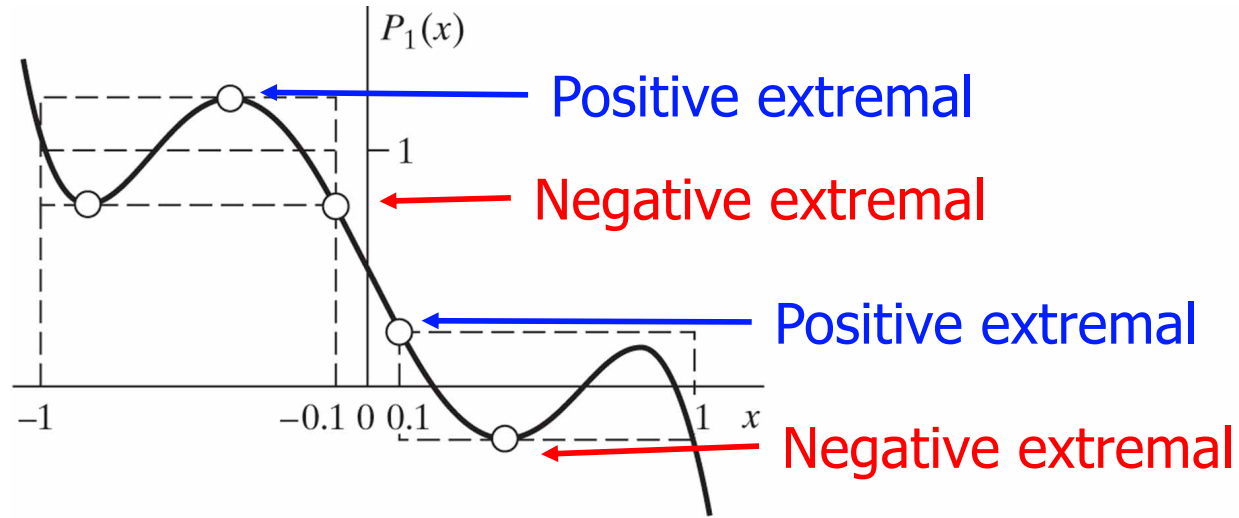
and the maximum error as

$$\|E(e^{j\omega})\|_{\infty} = \operatorname{argmax}_{x \in \Omega} |E(e^{j\omega})|$$

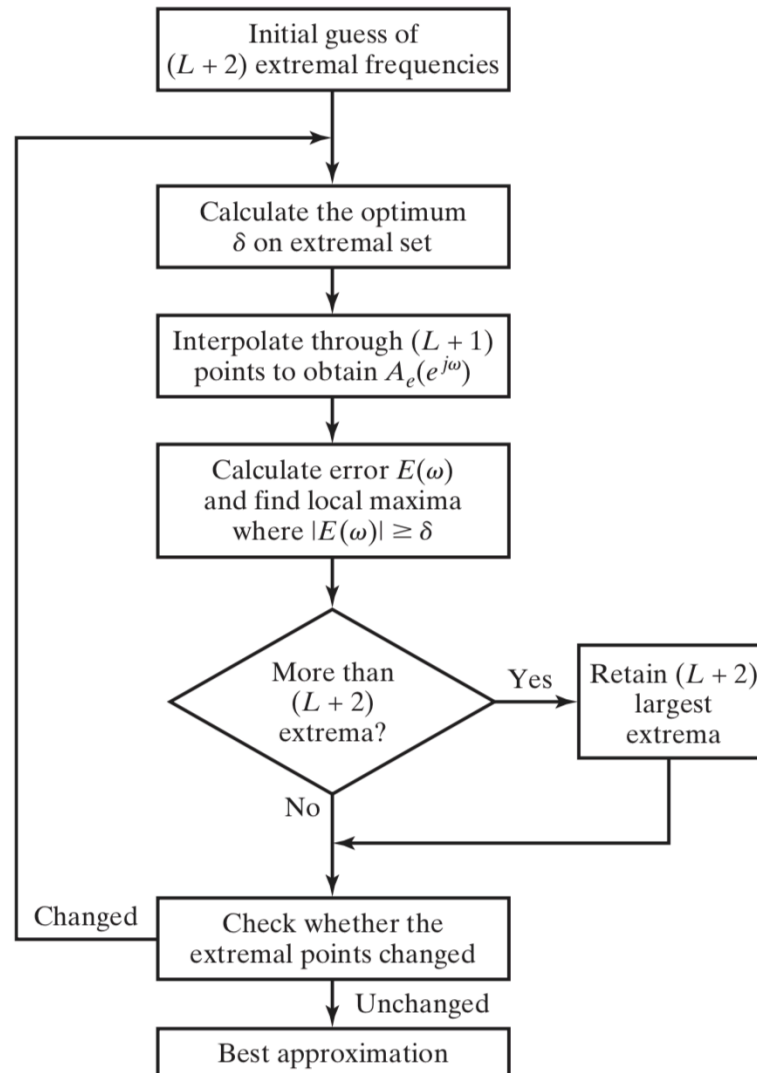
A necessary and sufficient condition that  $H(e^{j\omega})$  is the unique  $L$ th-order polynomial minimizing  $\|E(e^{j\omega})\|_{\infty}$  is that  $E(e^{j\omega})$  exhibit at least  $L + 2$  extremal frequencies, or “alternations”, that is there must exist at least  $L + 2$  values of  $\omega$ ,  $\omega_k \in \Omega$ ,  $k = [0, 1, \dots, L + 1]$ , such that  $\omega_0 < \omega_1 < \dots < \omega_{L+1}$ , and such that

$$E(e^{j\omega_k}) = -E(e^{j\omega_{k+1}}) = \pm (\|E(e^{j\omega})\|_{\infty}).$$

# Alternation Theorem Example – 5<sup>th</sup> order



# Parks–McClellan algorithm





# Min-Max Filter Design

---

## □ Constraints:

- min-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

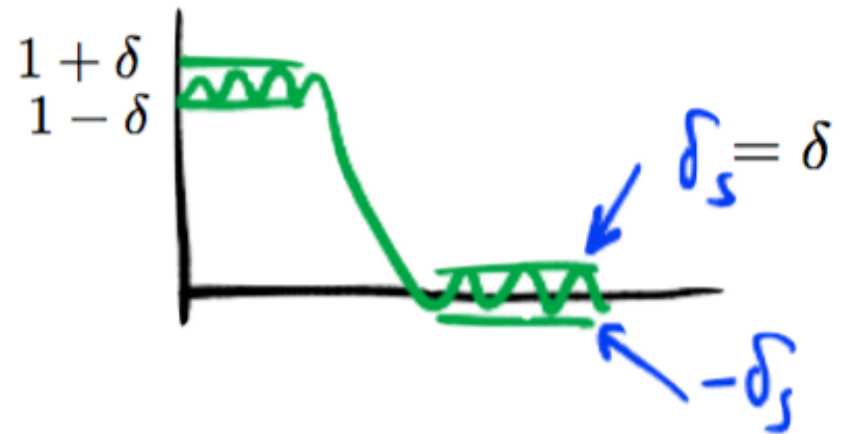
$$|H(e^{j\omega})| \leq \delta_s, \quad \omega_s \leq \omega \leq \pi$$

# Min-Max Ripple Design

- Given  $\omega_p$ ,  $\omega_s$ ,  $M$ , find  $\delta$ ,  $\tilde{h}_+$

minimize  $\delta$   
 Subject to :

$$\begin{aligned} 1 - \delta &\leq \tilde{H}(e^{j\omega_k}) \leq 1 + \delta & 0 \leq \omega_k \leq \omega_p \\ -\delta &\leq \tilde{H}(e^{j\omega_k}) \leq \delta & \omega_s \leq \omega_k \leq \pi \\ \delta &> 0 \end{aligned}$$



- Formulation is a linear program with solution  $\delta$ ,  $\tilde{h}_+$
- A well studied class of problems with good solvers



# Min-Max Ripple via LP

minimize  $\delta$   
subject to :

$$\begin{aligned} 1 - \delta &\preceq A_p \tilde{h}_+ \preceq 1 + \delta \\ -\delta &\preceq A_s \tilde{h}_+ \preceq \delta \\ \delta &> 0 \end{aligned}$$

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

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

capital P



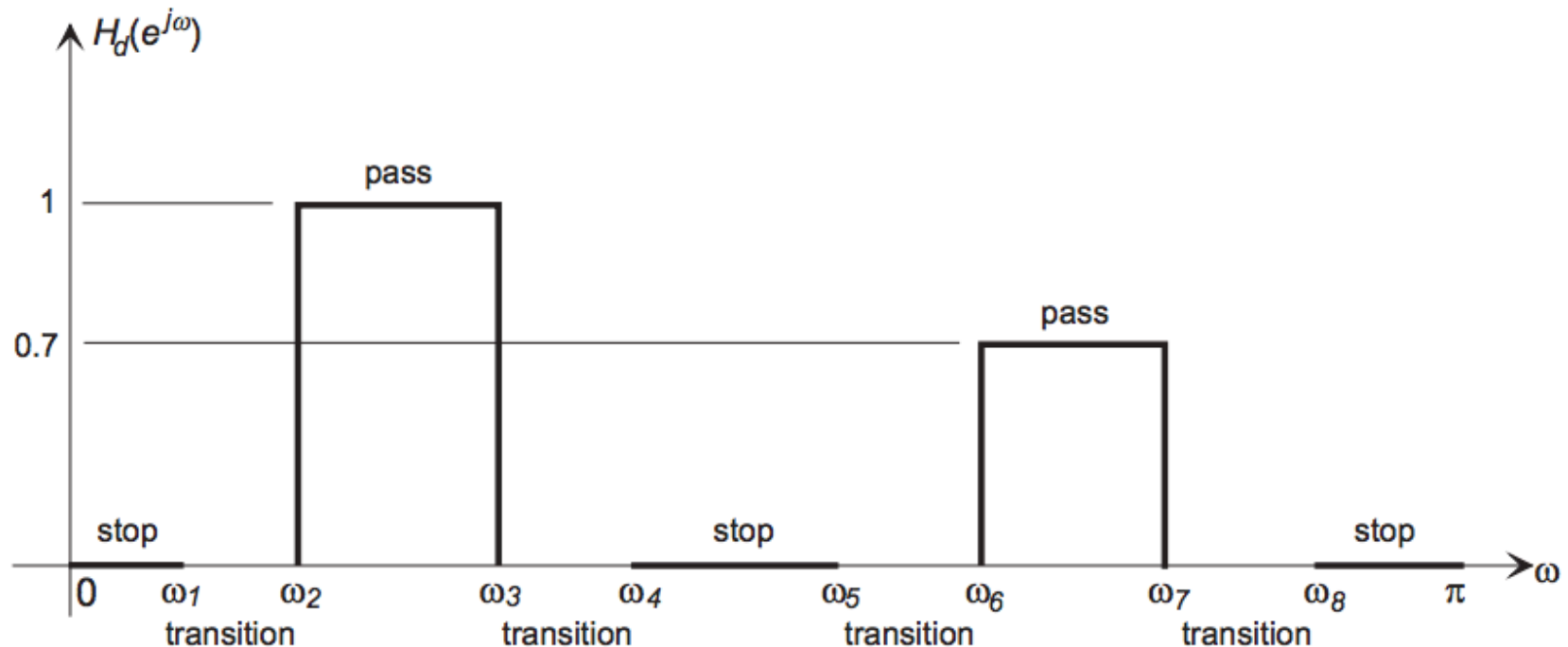
# MATLAB Parks-McClellan Function

---

□ `b = firpm(M, F, A, W)`

- `b` is the array of filter coefficients (impulse response)
- `M` is the filter order (`M+1` is the length of the filter),
- `F` is a vector of band edge frequencies in ascending order
- `A` is a set of filter gains at the band edges
- `W` is an optional set of relative weights to be applied to each of the bands

# MATLAB Parks-McClellan Function



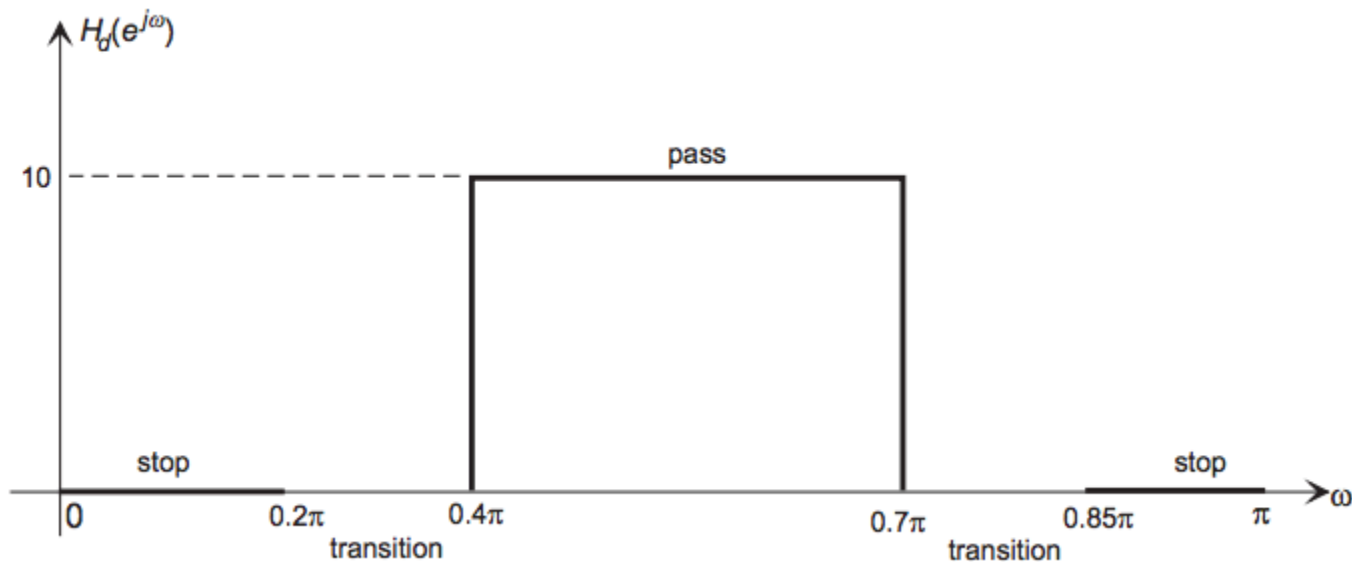
$$F = [0 \quad \omega_1 \quad \omega_2 \quad \omega_3 \quad \omega_4 \quad \omega_5 \quad \omega_6 \quad \omega_7 \quad \omega_8 \quad 1]$$

$$A = [0 \quad 0 \quad 1 \quad 1 \quad 0 \quad 0 \quad 0.7 \quad 0.7 \quad 0 \quad 0]$$

$$W = [10 \quad 1 \quad 10 \quad 1 \quad 10]$$

# MATLAB Example

- Design a 33 length PM filter and weight the stop-band ripple 10x more than the pass-band ripple



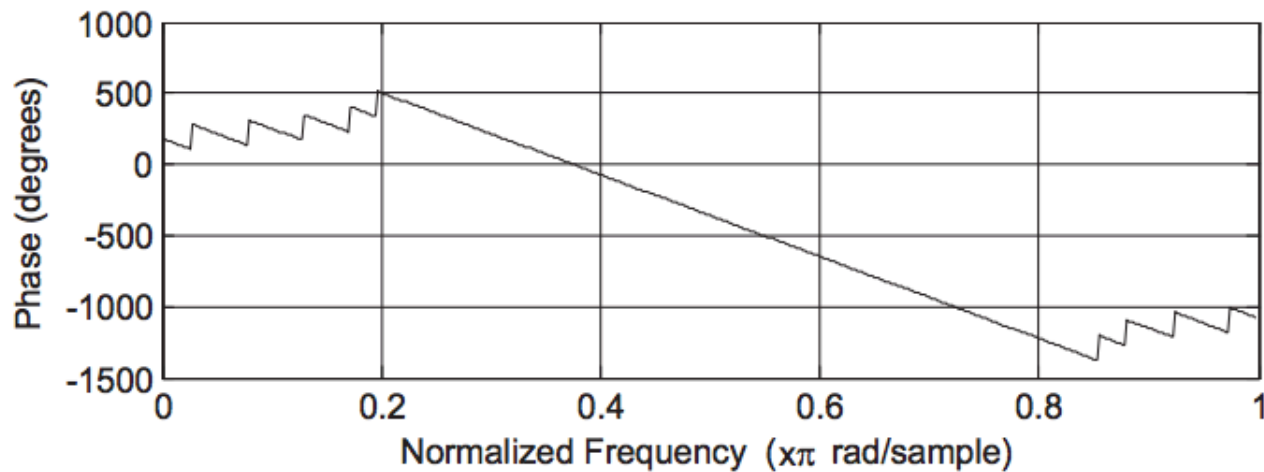
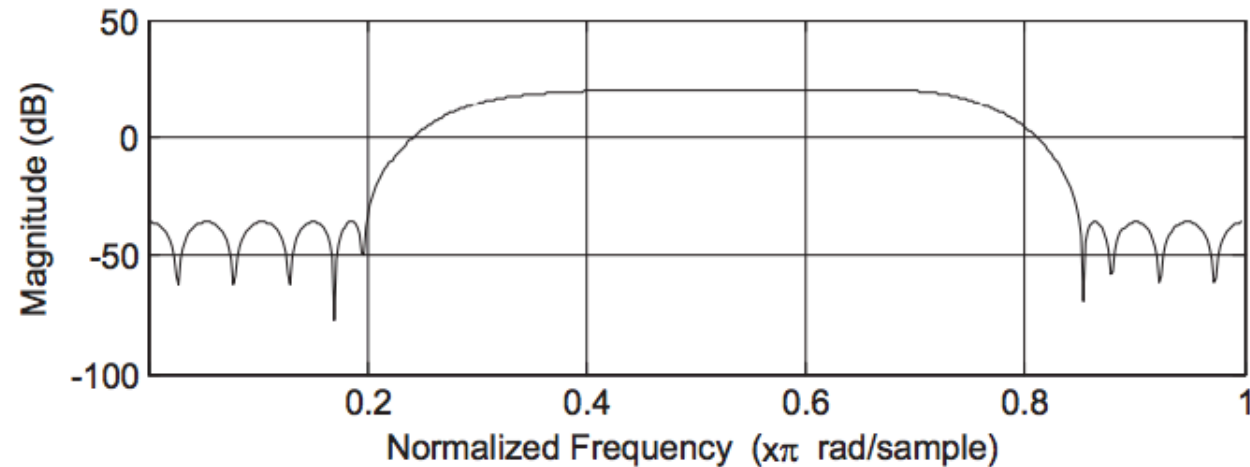
```
h=firpm(32,[0 0.2 0.4 0.7 0.85 1],[0 0 10 10 0 0],[10 1 10])  
freqz(h,1)
```



# MATLAB Example

Des  
ban

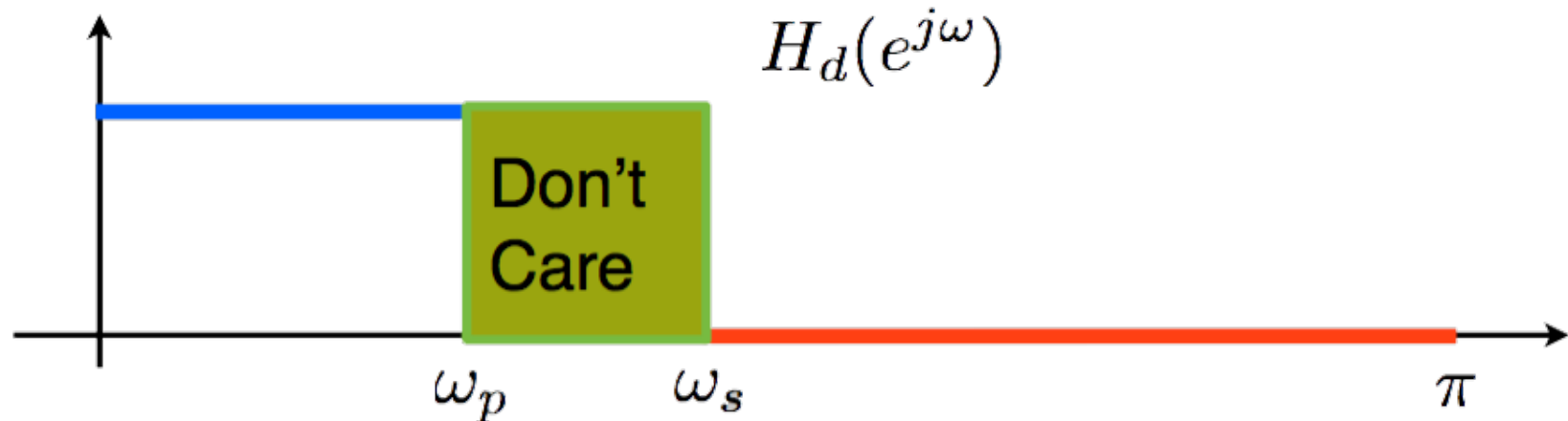
op-



1 10])

```
h=firp  
freqz(
```

# Optimality – Least Squares



□ Least Squares:

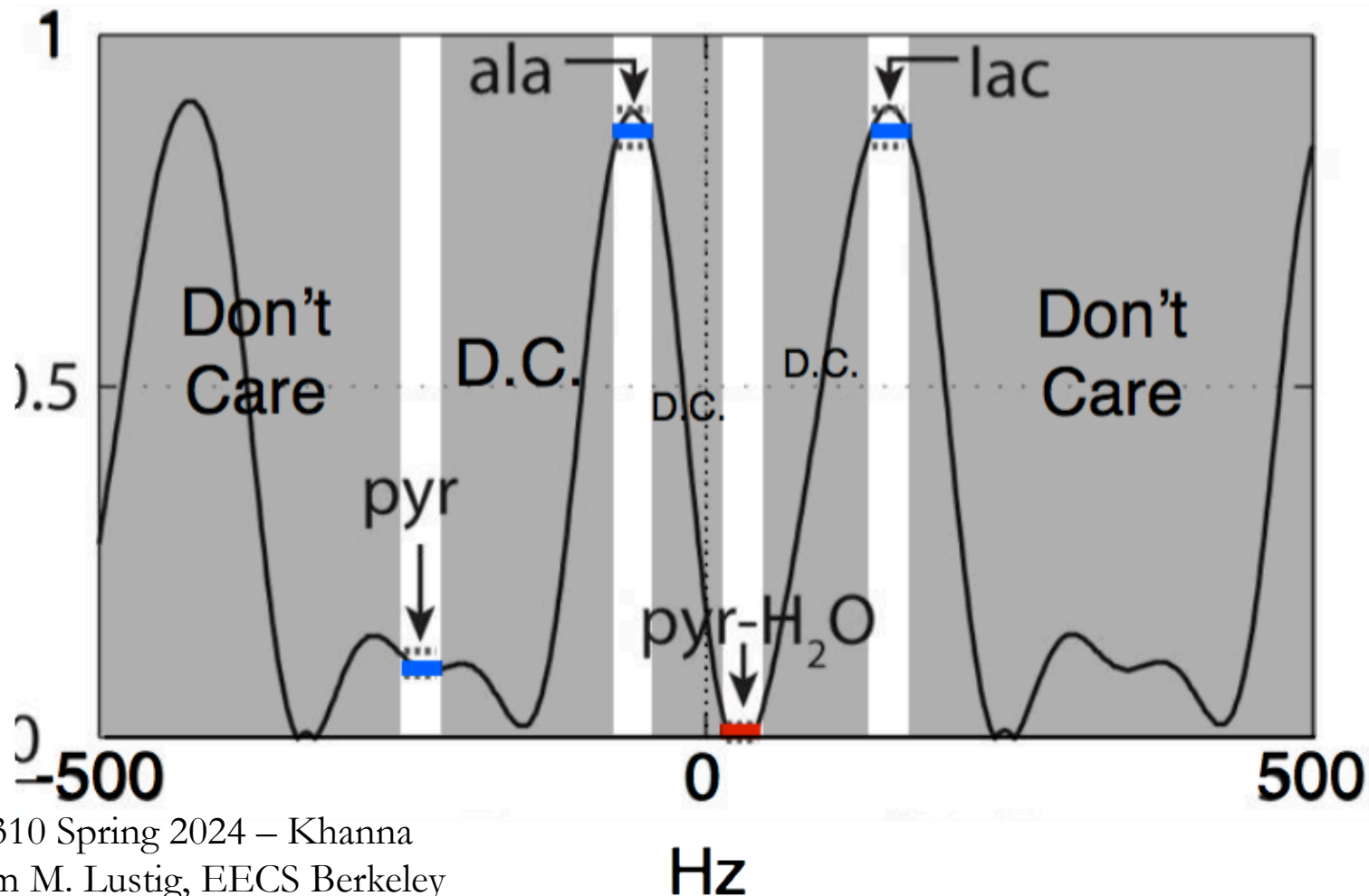
$$\text{minimize} \int_{\omega \in \text{care}} |H(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

□ Parks-McClellan

$$\min_{\{h_e[n]: 0 \leq n \leq L\}} \left( \max_{\omega \in F} |E(\omega)| \right),$$

# Example of Complex Filter

- ❑ Larson et. al, “Multiband Excitation Pulses for Hyperpolarized  $^{13}\text{C}$  Dynamic Chemical Shift Imaging” JMR 2008;194(1):121-127
- ❑ Need to design length 11 filter with following frequency response:





# Convex Optimization

---

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



# Using CVX (in Matlab)

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

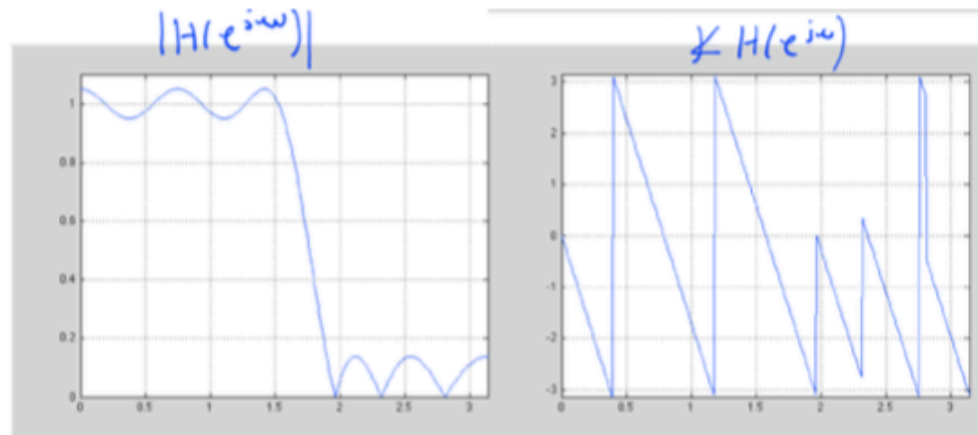
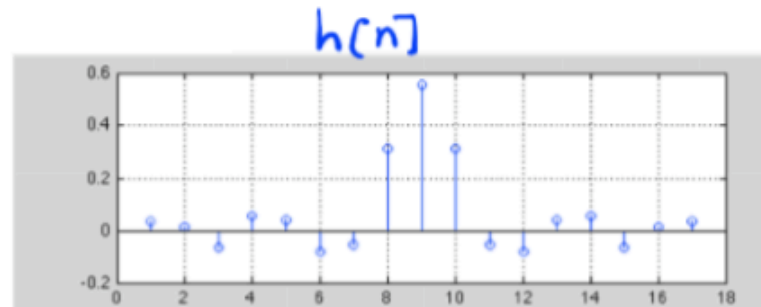
```
idxp = find(w <= wp);
idxs = find(w >= ws);
```

```
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)];
```





# Admin

---

- ❑ Project 1 - Due today @midnight
- ❑ Project 1 Submission
  - 2 assignments on Gradescope
  - Project 1 Report
    - Late penalty judged off this
  - Project 1 code
    - Can submit multiple files any format (.m, .mlx, .zip)