

▷ **9: Optimal IIR  
Design**

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**Error choices**

**Linear Least Squares**

**Frequency Sampling**

**Iterative Solution**

**Newton-Raphson**

**Magnitude-only**

**Specification**

**Hilbert Relations**

**Magnitude  $\leftrightarrow$  Phase  
Relation**

**Summary**

**MATLAB routines**

# 9: Optimal IIR Design

# Error choices

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### ▷ Error choices

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Specification

Hilbert Relations

Magnitude ↔ Phase Relation

Summary

MATLAB routines

We want to find a filter  $H(e^{j\omega}) = \frac{B(e^{j\omega})}{A(e^{j\omega})}$  that approximates a target response  $D(\omega)$ . Assume  $A$  is order  $N$  and  $B$  is order  $M$ .

Two possible error measures:

**Solution Error:**  $E_S(\omega) = W_S(\omega) \left( \frac{B(e^{j\omega})}{A(e^{j\omega})} - D(\omega) \right)$  *nonlinear*

**Equation Error:**  $E_E(\omega) = W_E(\omega) (B(e^{j\omega}) - D(\omega)A(e^{j\omega}))$  *linear*

We may know  $D(\omega)$  completely or else only  $|D(\omega)|$

We minimize  $\int_{-\pi}^{\pi} |E_*(\omega)|^p d\omega$

where  $p = 2$  (least squares) or  $\infty$  (minimax).

**Weight functions**  $W_*(\omega)$  are chosen to control relative errors at different frequencies.  $W_S(\omega) = |D(\omega)|^{-1}$  gives constant dB error.

We actually want to minimize  $E_S$  but  $E_E$  is easier because it gives rise to linear equations.

However if  $W_E(\omega) = \frac{W_S(\omega)}{|A(e^{j\omega})|}$ , then  $|E_E(\omega)| = |E_S(\omega)|$

# Linear Least Squares

## 9: Optimal IIR Design

### Error choices

#### Linear Least Squares

#### Frequency Sampling

#### Iterative Solution

#### Newton-Raphson

#### Magnitude-only

#### Specification

#### Hilbert Relations

#### Magnitude $\leftrightarrow$ Phase Relation

#### Summary

#### MATLAB routines

Overdetermined set of equations  $\mathbf{Ax} = \mathbf{b}$  (#equations  $>$  #unknowns)

We want to minimize  $\|\mathbf{e}\|^2$  where  $\mathbf{e} = \mathbf{Ax} - \mathbf{b}$

$$\|\mathbf{e}\|^2 = \mathbf{e}^T \mathbf{e} = (\mathbf{x}^T \mathbf{A}^T - \mathbf{b}^T) (\mathbf{Ax} - \mathbf{b})$$

Differentiate with respect to  $\mathbf{x}$ :

$$d(\mathbf{e}^T \mathbf{e}) = d\mathbf{x}^T \mathbf{A}^T (\mathbf{Ax} - \mathbf{b}) + (\mathbf{x}^T \mathbf{A}^T - \mathbf{b}^T) \mathbf{A} d\mathbf{x}$$

[since  $d(\mathbf{u}\mathbf{v}) = d\mathbf{u} \mathbf{v} + \mathbf{u} d\mathbf{v}$ ]

[since  $\mathbf{u}^T \mathbf{v} = \mathbf{v}^T \mathbf{u}$ ]

$$= 2d\mathbf{x}^T \mathbf{A}^T (\mathbf{Ax} - \mathbf{b})$$

$$= 2d\mathbf{x}^T (\mathbf{A}^T \mathbf{Ax} - \mathbf{A}^T \mathbf{b})$$

This is zero for any  $d\mathbf{x}$  iff  $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$

Thus  $\|\mathbf{e}\|^2$  is minimized if  $\mathbf{x} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$

These are the **Normal Equations** ("Normal" because  $\mathbf{A}^T \mathbf{e} = 0$ )

The **pseudoinverse**  $\mathbf{x} = \mathbf{A}^+ \mathbf{b}$  works even if  $\mathbf{A}^T \mathbf{A}$  is singular and finds the  $\mathbf{x}$  with minimum  $\|\mathbf{x}\|^2$  that minimizes  $\|\mathbf{e}\|^2$ .

This is a very widely used technique.

# Frequency Sampling

## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency

### ▷ Sampling

### Iterative Solution

### Newton-Raphson

### Magnitude-only

### Specification

### Hilbert Relations

### Magnitude ↔ Phase Relation

### Summary

### MATLAB routines

For every  $\omega$  we want:  $0 = W(\omega) (B(e^{j\omega}) - D(\omega)A(e^{j\omega}))$

$$= W(\omega) \left( \sum_{m=0}^M b[m]e^{-jm\omega} - D(\omega) \left( 1 + \sum_{n=1}^N a[n]e^{-jn\omega} \right) \right)$$

$$\Rightarrow \begin{pmatrix} \mathbf{u}(\omega)^T & \mathbf{v}(\omega)^T \end{pmatrix} \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix} = W(\omega)D(\omega)$$

where  $\mathbf{u}(\omega)^T = -W(\omega)D(\omega) \begin{bmatrix} e^{-j\omega} & e^{-j2\omega} & \dots & e^{-jN\omega} \end{bmatrix}$

$$\mathbf{v}(\omega)^T = W(\omega) \begin{bmatrix} 1 & e^{-j\omega} & e^{-j2\omega} & \dots & e^{-jM\omega} \end{bmatrix}$$

Choose  $K$  values of  $\omega$ ,  $\{ \omega_1 \dots \omega_K \}$  [with  $K \geq \frac{M+N+1}{2}$ ]

$$\begin{pmatrix} \mathbf{U}^T & \mathbf{V}^T \end{pmatrix} \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix} = \mathbf{d} \quad [K \text{ equations, } M + N + 1 \text{ unknowns}]$$

where  $\mathbf{U} = \begin{bmatrix} \mathbf{u}(\omega_1) & \dots & \mathbf{u}(\omega_K) \end{bmatrix},$

$$\mathbf{V} = \begin{bmatrix} \mathbf{v}(\omega_1) & \dots & \mathbf{v}(\omega_K) \end{bmatrix},$$

$$\mathbf{d} = \begin{bmatrix} W(\omega_1)D(\omega_1) & \dots & W(\omega_K)D(\omega_K) \end{bmatrix}^T$$

We want to force  $\mathbf{a}$  and  $\mathbf{b}$  to be real; find least squares solution to

$$\begin{pmatrix} \Re(\mathbf{U}^T) & \Re(\mathbf{V}^T) \\ \Im(\mathbf{U}^T) & \Im(\mathbf{V}^T) \end{pmatrix} \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix} = \begin{pmatrix} \Re(\mathbf{d}) \\ \Im(\mathbf{d}) \end{pmatrix}$$

# Iterative Solution

## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency Sampling

### ▷ Iterative Solution

### Newton-Raphson

### Magnitude-only

### Specification

### Hilbert Relations

### Magnitude ↔ Phase Relation

### Summary

### MATLAB routines

Least squares solution minimizes the  $E_E$  rather than  $E_S$ .

However  $E_E = E_S$  if  $W_E(\omega) = \frac{W_S(\omega)}{|A(e^{j\omega})|}$ .

We can use an iterative solution technique:

1 Select  $K$  frequencies  $\{\omega_k\}$  (e.g. uniformly spaced)

2 Initialize  $W_E(\omega_k) = W_S(\omega_k)$

3 Find least squares solution to

$$W_E(\omega_k) (B(e^{j\omega_k}) - D(\omega_k)A(e^{j\omega_k})) = 0 \forall k$$

4 Force  $A(z)$  to be stable

Replace pole  $p_i$  by  $(p_i^*)^{-1}$  whenever  $|p_i| \geq 1$

5 Update weights:  $W_E(\omega_k) = \frac{W_S(\omega_k)}{|A(e^{j\omega_k})|}$

6 Return to step 3 until convergence

But for faster convergence use Newton-Raphson ...

# Newton-Raphson

## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency Sampling

### Iterative Solution

### ▷ Newton-Raphson

### Magnitude-only

### Specification

### Hilbert Relations

### Magnitude ↔ Phase Relation

### Summary

### MATLAB routines

Newton: To solve  $f(x) = 0$  given an initial guess  $x_0$ , we write

$$f(x) \approx f(x_0) + (x - x_0)f'(x_0) \Rightarrow x = x_0 - \frac{f(x_0)}{f'(x_0)}$$

Converges very rapidly once  $x_0$  is close to the solution

So for each  $\omega_k$ , we can write (omitting the  $\omega$  and  $e^{j\omega}$  arguments)

$$\begin{aligned} E_S &\approx W_S \left( \frac{B_0}{A_0} - D \right) + \frac{W_S}{A_0} (B - B_0) - \frac{W_S B_0}{A_0^2} (A - A_0) \\ &= \frac{W_S}{A_0} \left( B_0 - A_0 D + B - B_0 - \frac{B_0}{A_0} (A - 1) - \frac{B_0}{A_0} + B_0 \right) \end{aligned}$$

From which we get a linear equation for each  $\omega_k$  :

$$\left( \begin{array}{cc} \frac{B_0}{DA_0} \mathbf{u}^T & \mathbf{v}^T \end{array} \right) \left( \begin{array}{c} \mathbf{a} \\ \mathbf{b} \end{array} \right) = W \left( A_0 D + \frac{B_0}{A_0} - B_0 \right)$$

where  $W = \frac{W_S}{A_0}$  and, as before,  $u_n(\omega) = -W(\omega)D(\omega)e^{-jn\omega}$

for  $n \in 1 : N$  and  $v_m(\omega) = W(\omega)e^{-jm\omega}$  for  $m \in 0 : M$ .

At each iteration, calculate  $A_0(e^{j\omega_k})$  and  $B_0(e^{j\omega_k})$  based on  $\mathbf{a}$  and  $\mathbf{b}$  from the previous iteration.

Then use linear least squares to minimize the linearized  $E_S$  using the above equation replicated for each of the  $\omega_k$ .

# Magnitude-only Specification

## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency Sampling

### Iterative Solution

### Newton-Raphson

### ▷ Magnitude-only

### Specification

### Hilbert Relations

### Magnitude $\leftrightarrow$ Phase Relation

### Summary

### MATLAB routines

If the filter specification only dictates the target magnitude:  $|D(\omega)|$ , we need to select the target phase.

### Solution:

Make an initial guess of the phase and then at each iteration update  $\angle D(\omega) = \angle \frac{B(e^{j\omega})}{A(e^{j\omega})}$ .

### Initial Guess:

If  $H(e^{j\omega})$  is **causal** and **minimum phase** then the magnitude and phase are not independent:

$$\begin{aligned}\angle H(e^{j\omega}) &= -\ln |H(e^{j\omega})| \circledast \cot \frac{\omega}{2} \\ \ln |H(e^{j\omega})| &= \ln |H(\infty)| + \angle H(e^{j\omega}) \circledast \cot \frac{\omega}{2}\end{aligned}$$

where  $\circledast$  is circular convolution and  $\cot x$  is taken to be zero for  $-\epsilon < x < \epsilon$  for some small value of  $\epsilon$  and we take the limit as  $\epsilon \rightarrow 0$ .

This result is a consequence of the **Hilbert Relations**.

# Hilbert Relations

## 9: Optimal IIR Design

Error choices

Linear Least Squares

Frequency Sampling

Iterative Solution

Newton-Raphson

Magnitude-only

Specification

▷ Hilbert Relations

Magnitude ↔ Phase Relation

Summary

MATLAB routines

We define  $t[n] = u[n-1] - u[-1-n]$

$$T(z) = \frac{z^{-1}}{1-z^{-1}} - \frac{z}{1-z} = \frac{1+z^{-1}}{1-z^{-1}}$$

$$\begin{aligned} T(e^{j\omega}) &= \frac{1+e^{-j\omega}}{1-e^{-j\omega}} = \frac{e^{j\frac{\omega}{2}} + e^{-j\frac{\omega}{2}}}{e^{j\frac{\omega}{2}} - e^{-j\frac{\omega}{2}}} \\ &= \frac{2 \cos \frac{\omega}{2}}{2j \sin \frac{\omega}{2}} = -j \cot \frac{\omega}{2} \end{aligned}$$

$$\begin{aligned} h[n] \rightarrow \text{even/odd parts: } h_e[n] &= \frac{1}{2} (h[n] + h[-n]) \\ h_o[n] &= \frac{1}{2} (h[n] - h[-n]) \end{aligned}$$

$$\begin{aligned} \text{so } \Re(H(e^{j\omega})) &= H_e(e^{j\omega}) \\ \Im(H(e^{j\omega})) &= -j H_o(e^{j\omega}) \end{aligned}$$

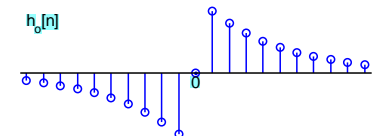
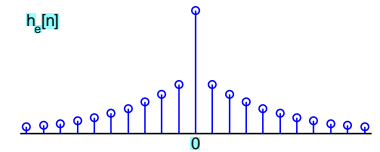
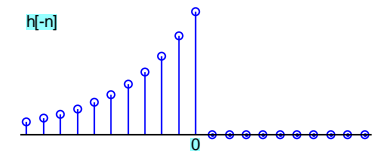
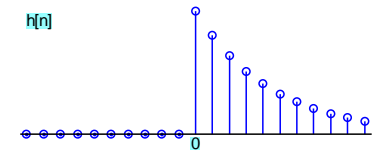
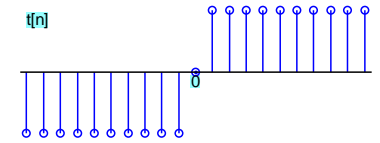
If  $h[n]$  is causal:  $h_o[n] = h_e[n]t[n]$

$$h_e[n] = h[0]\delta[n] + h_o[n]t[n]$$

Hence, for causal  $h[n]$ :

$$\begin{aligned} \Im(H(e^{j\omega})) &= -j (\Re(H(e^{j\omega})) \circledast -j \cot \frac{\omega}{2}) \\ &= -\Re(H(e^{j\omega})) \circledast \cot \frac{\omega}{2} \end{aligned}$$

$$\begin{aligned} \Re(H(e^{j\omega})) &= H(\infty) + j \Im(H(e^{j\omega})) \circledast -j \cot \frac{\omega}{2} \\ &= H(\infty) + \Im(H(e^{j\omega})) \circledast \cot \frac{\omega}{2} \end{aligned}$$





# Magnitude $\leftrightarrow$ Phase Relation

## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency Sampling

### Iterative Solution

### Newton-Raphson

### Magnitude-only Specification

### Hilbert Relations

### ▷ Magnitude $\leftrightarrow$

### Phase Relation

### Summary

### MATLAB routines

$$\text{Given } H(z) = g \frac{\prod (1 - q_m z^{-1})}{\prod (1 - p_n z^{-1})}$$

$$\begin{aligned} \ln H(z) &= \ln(g) + \sum \ln(1 - q_m z^{-1}) \\ &\quad - \sum \ln(1 - p_n z^{-1}) \\ &= \ln |H(z)| + j \angle H(z) \end{aligned}$$

Taylor Series:

$$\ln(1 - az^{-1}) = -az^{-1} - \frac{a^2}{2}z^{-2} - \frac{a^3}{3}z^{-3} - \dots$$

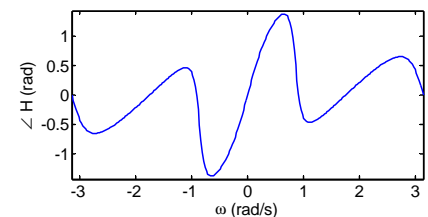
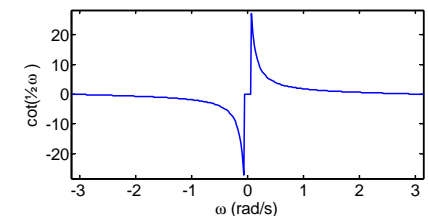
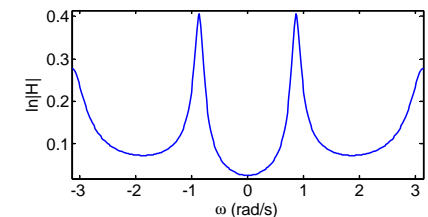
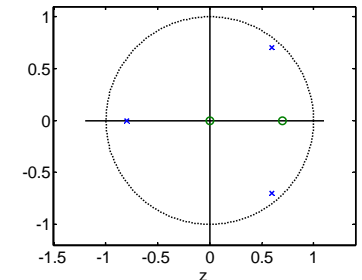
causal and stable provided  $|a| < 1$

So, if  $H(z)$  is **minimum phase** (all  $p_n$  and  $q_m$  inside unit circle) then  $\ln H(z)$  is the  $z$ -transform of a stable causal sequence and:

$$\begin{aligned} \angle H(e^{j\omega}) &= -\ln |H(e^{j\omega})| \circledast \cot \frac{\omega}{2} \\ \ln |H(e^{j\omega})| &= \ln |g| + \angle H(e^{j\omega}) \circledast \cot \frac{\omega}{2} \end{aligned}$$

Example:  $H(z) = \frac{10 - 7z^{-1}}{100 - 40z^{-1} - 11z^{-2} + 68z^{-3}}$

Note **symmetric dead band** in  $\cot \frac{\omega}{2}$  for  $|\omega| < \epsilon$



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## 9: Optimal IIR Design

### Error choices

### Linear Least Squares

### Frequency Sampling

### Iterative Solution

### Newton-Raphson

### Magnitude-only Specification

### Hilbert Relations

### Magnitude $\leftrightarrow$ Phase Relation

### ▷ Summary

### MATLAB routines

- Want to minimize solution error,  $E_S$ , but  $E_E$  gives linear equations:
  - $E_S(\omega) = W_S(\omega) \left( \frac{B(e^{j\omega})}{A(e^{j\omega})} - D(\omega) \right)$
  - $E_E(\omega) = W_E(\omega) (B(e^{j\omega}) - D(\omega)A(e^{j\omega}))$
  - use  $W_*(\omega)$  to weight errors at different  $\omega$ .
- Linear least squares: solution to overdetermined  $\mathbf{Ax} = \mathbf{b}$ 
  - Least squares error:  $\hat{\mathbf{x}} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$
- Closed form solution: least squares  $E_E$  at  $\{\omega_k\}$ 
  - use  $W_E(\omega) = \frac{W_S(\omega)}{|A(e^{j\omega})|}$  to approximate  $E_S$
  - use Taylor series to approximate  $E_S$  better (Newton-Raphson)
- Hilbert relations
  - relate  $\Re(H(e^{j\omega}))$  and  $\Im(H(e^{j\omega}))$  for causal stable sequences
  - $\Rightarrow$  relate  $\ln |H(e^{j\omega})|$  and  $\angle H(e^{j\omega})$  for causal stable minimum phase sequences

For further details see Mitra: 9.

# MATLAB routines

## 9: Optimal IIR Design

Error choices

Linear Least Squares

Frequency Sampling

Iterative Solution

Newton-Raphson

Magnitude-only

Specification

Hilbert Relations

Magnitude  $\leftrightarrow$  Phase  
Relation

Summary

▷ MATLAB routines

invfreqz

IIR design for complex response