Filter design

- FIR filters
- Chebychev design
- linear phase filter design
- equalizer design
- filter magnitude specifications

FIR filters

finite impulse response (FIR) filter:

$$y(t) = \sum_{\tau=0}^{n-1} h_{\tau} u(t-\tau), \quad t \in \mathbf{Z}$$

- (sequence) $u : \mathbf{Z} \to \mathbf{R}$ is input signal
- (sequence) $y : \mathbf{Z} \to \mathbf{R}$ is output signal
- h_i are called *filter coefficients*
- *n* is filter *order* or *length*

filter frequency response: $H : \mathbf{R} \to \mathbf{C}$

$$H(\omega) = h_0 + h_1 e^{-i\omega} + \dots + h_{n-1} e^{-i(n-1)\omega}$$
$$= \sum_{t=0}^{n-1} h_t \cos t\omega + i \sum_{t=0}^{n-1} h_t \sin t\omega$$

- (EE tradition uses $j = \sqrt{-1}$ instead of i)
- H is periodic and conjugate symmetric, so only need to know/specify for $0 \leq \omega \leq \pi$

FIR filter design problem: choose h so it and H satisfy/optimize specs

example: (lowpass) FIR filter, order n = 21

impulse response *h*:



frequency response magnitude (*i.e.*, $|H(\omega)|$):



frequency response phase (*i.e.*, $\angle H(\omega)$):



Filter design

Chebychev design

minimize $\max_{\omega \in [0,\pi]} |H(\omega) - H_{des}(\omega)|$

- *h* is optimization variable
- $H_{des} : \mathbf{R} \to \mathbf{C}$ is (given) desired transfer function
- convex problem
- can add constraints, e.g., $|h_i| \leq 1$

sample (discretize) frequency:

minimize
$$\max_{k=1,\ldots,m} |H(\omega_k) - H_{des}(\omega_k)|$$

- sample points $0 \le \omega_1 < \cdots < \omega_m \le \pi$ are fixed (e.g., $\omega_k = k\pi/m$)
- $m \gg n$ (common rule-of-thumb: m = 15n)
- yields approximation (relaxation) of problem above

Chebychev design via SOCP:

minimize
$$t$$

subject to $||A^{(k)}h - b^{(k)}|| \le t, \quad k = 1, \dots, m$

where

$$A^{(k)} = \begin{bmatrix} 1 & \cos \omega_k & \cdots & \cos(n-1)\omega_k \\ 0 & -\sin \omega_k & \cdots & -\sin(n-1)\omega_k \end{bmatrix}$$
$$b^{(k)} = \begin{bmatrix} \Re H_{des}(\omega_k) \\ \Im H_{des}(\omega_k) \end{bmatrix}$$
$$h = \begin{bmatrix} h_0 \\ \vdots \\ h_{n-1} \end{bmatrix}$$

Linear phase filters

suppose

- n = 2N + 1 is odd
- impulse response is symmetric about midpoint:

$$h_t = h_{n-1-t}, \quad t = 0, \dots, n-1$$

then

$$H(\omega) = h_0 + h_1 e^{-i\omega} + \dots + h_{n-1} e^{-i(n-1)\omega}$$

= $e^{-iN\omega} (2h_0 \cos N\omega + 2h_1 \cos(N-1)\omega + \dots + h_N)$
 $\stackrel{\Delta}{=} e^{-iN\omega} \widetilde{H}(\omega)$

- term $e^{-iN\omega}$ represents N-sample delay
- $\widetilde{H}(\omega)$ is real
- $|H(\omega)| = |\widetilde{H}(\omega)|$
- called **linear phase** filter ($\angle H(\omega)$ is linear except for jumps of $\pm \pi$)

Lowpass filter specifications



idea:

- pass frequencies in *passband* $[0, \omega_{\rm p}]$
- block frequencies in *stopband* $[\omega_{\rm s},\pi]$

specifications:

• maximum passband ripple ($\pm 20 \log_{10} \delta_1$ in dB):

$$1/\delta_1 \le |H(\omega)| \le \delta_1, \quad 0 \le \omega \le \omega_p$$

• minimum stopband attenuation $(-20 \log_{10} \delta_2 \text{ in dB})$:

$$|H(\omega)| \le \delta_2, \quad \omega_{\rm s} \le \omega \le \pi$$

Linear phase lowpass filter design

- sample frequency
- can assume wlog $\widetilde{H}(0) > 0$, so ripple spec is

$$1/\delta_1 \le \widetilde{H}(\omega_k) \le \delta_1$$

design for maximum stopband attenuation:

$$\begin{array}{ll} \mbox{minimize} & \delta_2 \\ \mbox{subject to} & 1/\delta_1 \leq \widetilde{H}(\omega_k) \leq \delta_1, & 0 \leq \omega_k \leq \omega_p \\ & -\delta_2 \leq \widetilde{H}(\omega_k) \leq \delta_2, & \omega_{\rm s} \leq \omega_k \leq \pi \end{array}$$

- passband ripple δ_1 is given
- an LP in variables h, δ_2
- known (and used) since 1960's
- can add other constraints, e.g., $|h_i| \leq \alpha$

variations and extensions:

- fix δ_2 , minimize δ_1 (convex, but not LP)
- fix δ_1 and δ_2 , minimize ω_s (quasiconvex)
- fix δ_1 and δ_2 , minimize order n (quasiconvex)

example

- linear phase filter, n = 21
- passband $[0, 0.12\pi]$; stopband $[0.24\pi, \pi]$
- max ripple $\delta_1 = 1.012 \ (\pm 0.1 \text{dB})$
- design for maximum stopband attenuation

impulse response *h*:



frequency response magnitude (*i.e.*, $|H(\omega)|$):



frequency response phase (*i.e.*, $\angle H(\omega)$):



Filter design

Equalizer design



equalization: given

- G (unequalized frequency response)
- G_{des} (desired frequency response)

design (FIR equalizer) H so that $\widetilde{G} \stackrel{\Delta}{=} GH \approx G_{des}$

- common choice: $G_{des}(\omega) = e^{-iD\omega}$ (delay) *i.e.*, equalization is deconvolution (up to delay)
- can add constraints on H, e.g., limits on $|h_i|$ or $\max_{\omega} |H(\omega)|$

Chebychev equalizer design:

minimize
$$\max_{\omega \in [0,\pi]} \left| \widetilde{G}(\omega) - G_{des}(\omega) \right|$$

convex; SOCP after sampling frequency

time-domain equalization: optimize impulse response \tilde{g} of equalized system e.g., with $G_{des}(\omega) = e^{-iD\omega}$,

$$g_{\rm des}(t) = \begin{cases} 1 & t = D\\ 0 & t \neq D \end{cases}$$

sample design:

minimize $\max_{t \neq D} |\tilde{g}(t)|$ subject to $\tilde{g}(D) = 1$

- an LP
- can use $\sum_{t \neq D} \tilde{g}(t)^2$ or $\sum_{t \neq D} |\tilde{g}(t)|$

extensions:

- can impose (convex) constraints
- can mix time- and frequency-domain specifications
- can equalize multiple systems, i.e., choose H so

$$G^{(k)}H \approx G_{\text{des}}, \quad k = 1, \dots, K$$

- can equalize multi-input multi-output systems (*i.e.*, G and H are matrices)
- extends to multidimensional systems, *e.g.*, image processing

Equalizer design example

unequalized system G is 10th order FIR:





design $30 {\rm th}$ order FIR equalizer with $\widetilde{G}(\omega) \approx e^{-i 10 \omega}$

Filter design

Chebychev equalizer design:

minimize
$$\max_{\omega} \left| \tilde{G}(\omega) - e^{-i10\omega} \right|$$





time-domain equalizer design:

minimize $\max_{t \neq 10} |\tilde{g}(t)|$





Filter magnitude specifications

transfer function *magnitude spec* has form

$$L(\omega) \le |H(\omega)| \le U(\omega), \quad \omega \in [0,\pi]$$

where $L, U : \mathbf{R} \to \mathbf{R}_+$ are given

- lower bound is **not** convex in filter coefficients *h*
- arises in many applications, e.g., audio, spectrum shaping
- can change variables to solve via convex optimization

Autocorrelation coefficients

autocorrelation coefficients associated with impulse response $h = (h_0, \ldots, h_{n-1}) \in \mathbf{R}^n$ are

$$r_t = \sum_{\tau} h_{\tau} h_{\tau+t}$$

(we take $h_k = 0$ for k < 0 or $k \ge n$)

- $r_t = r_{-t}; r_t = 0 \text{ for } |t| \ge n$
- hence suffices to specify $r = (r_0, \ldots, r_{n-1}) \in \mathbf{R}^n$

Fourier transform of autocorrelation coefficients is

$$R(\omega) = \sum_{\tau} e^{-i\omega\tau} r_{\tau} = r_0 + \sum_{t=1}^{n-1} 2r_t \cos \omega t = |H(\omega)|^2$$

- always have $R(\omega) \geq 0$ for all ω

• can express magnitude specification as

$$L(\omega)^2 \le R(\omega) \le U(\omega)^2, \quad \omega \in [0,\pi]$$

 \ldots convex in r

Spectral factorization

question: when is $r \in \mathbf{R}^n$ the autocorrelation coefficients of some $h \in \mathbf{R}^n$? answer: (spectral factorization theorem) if and only if $R(\omega) \ge 0$ for all ω

- spectral factorization condition is convex in \boldsymbol{r}
- many algorithms for spectral factorization, i.e., finding an h s.t. $R(\omega)=|H(\omega)|^2$

magnitude design via autocorrelation coefficients:

- use r as variable (instead of h)
- add spectral factorization condition $R(\omega) \geq 0$ for all ω
- optimize over r
- $\bullet\,$ use spectral factorization to recover h

log-Chebychev magnitude design

choose h to minimize

$$\max_{\omega} |20 \log_{10} |H(\omega)| - 20 \log_{10} D(\omega)|$$

- D is desired transfer function magnitude $(D(\omega) > 0 \text{ for all } \omega)$
- find minimax logarithmic (dB) fit

reformulate as

 $\begin{array}{ll} \mbox{minimize} & t \\ \mbox{subject to} & D(\omega)^2/t \leq R(\omega) \leq t D(\omega)^2, \quad 0 \leq \omega \leq \pi \\ \end{array}$

- convex in variables r, t
- constraint includes spectral factorization condition

example: 1/f (pink noise) filter (*i.e.*, $D(\omega) = 1/\sqrt{\omega}$), n = 50, log-Chebychev design over $0.01\pi \le \omega \le \pi$



optimal fit: $\pm 0.5 dB$

MIT OpenCourseWare http://ocw.mit.edu

6.079 / 6.975 Introduction to Convex Optimization Fall 2009

For information about citing these materials or our Terms of Use, visit: http://ocw.mit.edu/terms.