Nonnegative polynomials, SDP formulations, and primal-dual interior-point methods

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Outline

- SDP representation of nonnegative (trigonometric) polynomials
- primal-dual interior-point methods for SDP

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Nonnegative trigonometric polynomials

 $X(\omega) = x_0 + 2x_1 \cos \omega + \dots + 2x_n \cos n\omega \ge 0, \quad \omega \in [0,\pi]$

- an infinite set of linear inequalities in $x \in \mathbf{R}^{n+1}$
- defines a closed convex cone $C_{n+1} = \{x \mid X(\omega) \ge 0\}$

spectral factorization (Riesz-Fejér theorem)

 $x \in C_{n+1}$ if and only if there exist $y_0, \ldots, y_n \in \mathbf{R}$ such that

$$X(\omega) = |y_0 + y_1 e^{-j\omega} + y_2 e^{-2j\omega} + \dots + y_n e^{-nj\omega}|^2$$

LMI characterization (equality form)

 $x \in C_{n+1}$ if and only if there exists

$$Y = \begin{bmatrix} Y_{00} & Y_{10} & \cdots & Y_{n0} \\ Y_{10} & Y_{11} & \cdots & Y_{n1} \\ \vdots & \vdots & \ddots & \vdots \\ Y_{n0} & Y_{n1} & \cdots & Y_{nn} \end{bmatrix} \succeq 0$$

such that

$$x_0 = Y_{00} + Y_{11} + \dots + Y_{nn}$$

$$x_1 = Y_{10} + Y_{21} + \dots + Y_{n-1,n}$$

$$\vdots$$

$$x_n = Y_{n0}$$
i.e., $x_k = \operatorname{Tr}(E^k Y)$ where $E = \begin{bmatrix} 0 & I_n \\ 0 & 0 \end{bmatrix}$

proof

• if
$$x_k = \mathbf{Tr}(E^k Y)$$
 with $Y \succeq 0$, then

$$X(\omega) = \begin{bmatrix} 1\\ e^{j\omega}\\ \vdots\\ e^{nj\omega} \end{bmatrix}^{H} \begin{bmatrix} Y_{00} & Y_{10} & \cdots & Y_{n0}\\ Y_{10} & Y_{11} & \cdots & Y_{n1}\\ \vdots & \vdots & \ddots & \vdots\\ Y_{n0} & Y_{n1} & \cdots & Y_{nn} \end{bmatrix} \begin{bmatrix} 1\\ e^{j\omega}\\ \vdots\\ e^{nj\omega} \end{bmatrix} \ge 0$$

• if $X(\omega) \ge 0$, expanding $X(\omega) = |y_0 + y_1 e^{-j\omega} + \dots + y_n e^{-nj\omega}|^2$ gives

i.e., $x_k = y^T E^k y = \mathbf{Tr}(E^k y y^T)$

LMI characterization (inequality form)

$$\begin{array}{rcl}
x_{0} &=& Y_{00} + Y_{11} + \dots + Y_{nn} \\
x_{1} &=& Y_{10} + Y_{21} + \dots + Y_{n-1,n} \\
& \vdots \\
x_{n} &=& Y_{n0}
\end{array}$$

if and only if there exists exists $P \in \mathbf{S}^n$ such that

$$Y(x,P) = \begin{bmatrix} 0 & 0 \\ 0 & P \end{bmatrix} - \begin{bmatrix} P & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & \cdots & 0 & x_n \\ \vdots & \cdots & \vdots & \vdots \\ 0 & \cdots & 0 & x_1 \\ x_n & \cdots & x_1 & x_0 \end{bmatrix}$$

therefore, $x \in C_{n+1}$ if and only if $Y(x, P) \succeq 0$ for some $P \in \mathbf{S}^n$

Dual cone

$$C_{n+1}^* = \{ z \mid z^T x \ge 0 \text{ for all } x \in C_{n+1} \}$$

LMI characterization: $z \in C^*_{n+1}$ if and only if

$$Z = \begin{bmatrix} 2z_0 & z_1 & z_2 & \cdots & z_n \\ z_1 & 2z_0 & z_1 & \cdots & z_{n-1} \\ z_2 & z_1 & 2z_0 & \cdots & z_{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ z_n & z_{n-1} & z_{n-2} & \cdots & 2z_0 \end{bmatrix} \succeq 0$$

proof: $z^T x \ge 0$ for all $x \in C_{n+1}$ if for all y,

$$\sum_{k=0}^{n} z_k(y^T E^k y) = \frac{1}{2} y^T Z y \ge 0$$

Nonnegative real polynomials on [-1, 1]

change of variables $t = \cos \omega$ maps $[0, \pi]$ to [-1, 1] and $X(\omega)$ to

$$Q(t) = X(\cos^{-1} t) = x_0 p_0(t) + 2x_1 p_1(t) + \dots + 2x_n p_n(t)$$

where $p_k(t) = \cos(k \cos^{-1} t)$ (the kth Chebyshev polynomial)

SOS representation (for n = 2m): polynomial $Q(t) \ge 0$ on [-1, 1] iff $Q(t) = |y_0 + y_1 e^{-j\omega} + y_2 e^{-2j\omega} + \dots + y_n e^{-nj\omega}|^2 \Big|_{\omega = \cos^{-1} t}$ $= (y_m p_0(t) + (y_{m-1} + y_{m+1})p_1(t) + \dots + (y_0 + y_n)p_m(t))^2$ $+ (1 - t^2) ((y_{m-1} - y_{m+1})q_0(t) + \dots + (y_0 - y_n)q_{n-1}(t))^2$

where $q_{k-1}(t) = \sin(k \cos^{-1} t) / \sin(\cos^{-1} t)$ (Cheb. polyn. of 2nd kind)

 $a_i^T x$ are coefficients of

$$Q(t) = (a_0^T x)\tilde{p}_0(t) + (a_1^T x)\tilde{p}_1(t) + \dots + (a_n^T x)\tilde{p}_n(t)$$

in basis of shifted Chebyshev polynomials

$$\tilde{p}_k(t) = p_k((2t - (t_1 + t_2))/(t_2 - t_1))$$

Example: Magnitude FIR filter design magnitude constraints

$$L \le |h_0 + h_1 e^{-j\omega} + \dots + h_n e^{-nj\omega}| \le U, \quad \omega \in [\omega_1, \omega_2]$$

• $H(\omega) = \sum_k h_k e^{-kj\omega}$ is frequency response of FIR filter

• not convex in filter coefficients h_i

change of variables $x_k = \sum_{i=0}^{n-k} h_i h_{i+k}$ gives equivalent constraints

$$L^{2} \leq x_{0} + 2x_{1}\cos\omega + \dots + 2x_{n}\cos n\omega \leq U^{2}, \quad \omega \in [\omega_{1}, \omega_{2}]$$
$$x_{0} + 2x_{1}\cos\omega + \dots + 2x_{n}\cos n\omega \geq 0, \quad \omega \in [0, \pi]$$

- from x, obtain h by spectral factorization
- convex constraints in x, representable as LMIs

LMI formulation (auxiliary variables $P_1, P_2, P_3 \in \mathbf{S}^n$)

•
$$X(\omega) \ge 0$$
, $\omega \in [0, \pi]$:

$$\begin{bmatrix} 0 & 0 \\ 0 & P_1 \end{bmatrix} - \begin{bmatrix} P_1 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & \cdots & 0 & x_n \\ \vdots & \cdots & \vdots & \vdots \\ 0 & \cdots & 0 & x_1 \\ x_n & \cdots & x_1 & x_0 \end{bmatrix} \succeq 0$$

• $L^2 \leq X(\omega) \leq U^2$, $\omega \in [\omega_1, \omega_2]$:

$$\begin{bmatrix} 0 & 0 \\ 0 & P_2 \end{bmatrix} - \begin{bmatrix} P_2 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & \cdots & 0 & a_n^T x \\ \vdots & \cdots & \vdots & \vdots \\ 0 & \cdots & 0 & a_1^T x \\ a_n^T x & \cdots & a_1^T x & a_0^T x - L^2 \end{bmatrix} \succeq 0$$
$$\begin{bmatrix} 0 & 0 \\ 0 & P_3 \end{bmatrix} - \begin{bmatrix} P_3 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & \cdots & 0 & -a_n^T x \\ \vdots & \cdots & \vdots & \vdots \\ 0 & \cdots & 0 & -a_1^T x \\ -a_n^T x & \cdots & -a_1^T x & U^2 - a_0^T x \end{bmatrix} \succeq 0$$

example: peak-constrained least-squares filter with N bands $[\alpha_k, \beta_k]$

minimize
$$\int_{\text{stopbands}} |H(\omega)|^2 d\omega$$

subject to $L_k \leq |H(\omega)| \leq U_k$, $\omega \in [\alpha_k, \beta_k]$, $k = 1, \dots, N$

use
$$X(\omega) = |H(\omega)|^2 = x_0 + 2x_1 \cos \omega + \dots + 2x_n \cos n\omega$$
, to get
minimize $\int_{\text{stopbands}} X(\omega) d\omega$
subject to $L_k^2 \leq X(\omega) \leq U_k^2$, $\omega \in [\alpha_k, \beta_k]$, $k = 1, \dots, N$
 $X(\omega) \geq 0$, $\omega \in [0, \pi]$

- standard method: discretize constraints and solve an LP
- SDP method: solve SDP with variables x, 2N + 1 matrices $P_k \in \mathbf{S}^n$

general problem

minimize
$$q^T x$$

subject to $\begin{bmatrix} 0 & 0 \\ 0 & P_k \end{bmatrix} - \begin{bmatrix} P_k & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=0}^p x_i M_{ki} \succeq N_k, \quad k = 1, \dots, L$

- variables $x \in \mathbf{R}^p$, $P_k \in \mathbf{S}^n$
- P_k are auxiliary variables, introduced to formulate semi-infinite inequalities as LMIs
- expensive to solve using general-purpose SDP software

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Interior-point SDP methods

early methods (1990-1995)

- projective method, implemented in the Matlab LMI control toolbox
- potential reduction methods, implemented in SDPSOL
- barrier methods

more recent methods

- primal-dual path-following methods
- general-purpose software packages: Sedumi, SDPT3, SDPPACK, SDPA, CSDP, DSDP, Yalmip, . . .

SDP duality

primal SDP

 $\begin{array}{ll} \text{minimize} & \langle c,y\rangle \\ \text{subject to} & \mathcal{A}(y)+S=B, \quad S\succeq 0 \end{array}$

- variable $x \in \mathcal{V}$, $S \in \mathbf{S}^n$ (slack variable)
- \mathcal{A} is linear mapping from \mathcal{V} to \mathbf{S}^n

dual SDP

$$\begin{array}{ll} \mbox{maximize} & -\operatorname{\mathbf{Tr}}(BZ) \\ \mbox{subject to} & \mathcal{A}^{\mathrm{adj}}(Z) + c = 0, \quad Z \succeq 0 \end{array}$$

variable $Z \in \mathbf{S}^n$; $\mathcal{A}^{\mathsf{adj}} : \mathbf{S}^n \to \mathcal{V}$ is adjoint of \mathcal{A}

optimal values are equal (if primal or dual is strictly feasible)

Primal-dual path-following algorithm

(Tütüncü, Toh, Todd)

select starting point $S \succ 0$, $Z \succ 0$, any y; repeat the following steps

- 1. Verify stopping criteria.
- 2. Compute the Nesterov-Todd scaling matrix R: defined by

 $R^T S^{-1} R = \operatorname{diag}(\lambda)^{-1}, \qquad R^T Z R = \operatorname{diag}(\lambda), \qquad \lambda \in \mathbf{R}_{++}^n$

3. Compute affine scaling directions ΔZ^{a} , ΔS^{a} , Δy^{a} :

$$\begin{aligned} \mathcal{H}(\Delta Z^{\mathsf{a}}S + Z\Delta S^{\mathsf{a}}) &= -\operatorname{diag}(\lambda)^{2} \\ \Delta S^{\mathsf{a}} + \mathcal{A}(\Delta y^{\mathsf{a}}) &= -(\mathcal{A}(y) + S - B) \\ \mathcal{A}^{\mathsf{adj}}(\Delta Z^{\mathsf{a}}) &= -(\mathcal{A}^{\mathsf{adj}}(Z) + c) \end{aligned}$$

where $\mathcal{H}(X) = \frac{1}{2}(R^TXR^{-T} + R^{-1}X^TR)$

4. Compute centering-corrector steps ΔZ^{c} , ΔS^{c} , Δy^{c} :

$$\mathcal{H}(\Delta Z^{\mathsf{c}}S + Z\Delta S^{\mathsf{c}}) = \rho I - \mathcal{H}(\Delta Z^{\mathsf{a}}\Delta S^{\mathsf{a}})$$
$$\Delta S^{\mathsf{c}} + \mathcal{A}(\Delta y^{\mathsf{c}}) = 0$$
$$\mathcal{A}^{\mathsf{adj}}(\Delta Z^{\mathsf{c}}) = 0$$

with ρ calculated based on ${\rm Tr}(SZ)$, $\Delta Z^{\rm a}$, $\Delta S^{\rm a}$

5. Update primal and dual iterates:

$$y := y + \alpha \Delta y, \qquad S := S + \alpha \Delta S, \qquad Z := Z + \beta \Delta Z$$

where $\Delta y = \Delta y^{a} + \Delta y^{c}$, $\Delta S = \Delta S^{a} + \Delta S^{c}$, $\Delta Z = \Delta Z^{a} + \Delta Z^{c}$,

$$\alpha = \min\{1, 0.99 \sup\{\alpha \mid S + \alpha \Delta S \succeq 0\}\}$$

$$\beta = \min\{1, 0.99 \sup\{\beta \mid Z + \beta \Delta Z \succeq 0\}\}$$

Overall complexity

- number of iterations is small (< 30)
- at each iteration, solve two sets of equations ('Newton equations')

$$\begin{aligned} \mathcal{H}(\Delta ZS + Z\Delta S) &= D_1 \\ \Delta S + \mathcal{A}(\Delta y) &= D_2 \\ \mathcal{A}^{\mathsf{adj}}(\Delta Z) &= d \end{aligned}$$

where

$$\mathcal{H}(X) = \frac{1}{2} (R^T X R^{-T} + R^{-1} X^T R)$$

values of R (NT scaling matrix), D_1 , D_2 , d change at each iteration

• equations for other primal-dual methods are similar (with different R)

General-purpose implementation

• eliminate ΔS from $\mathcal{H}(\Delta ZS + Z\Delta S) = D_1$:

$$-W\Delta ZW + \mathcal{A}(\Delta y) = D$$
(1)
$$\mathcal{A}^{\mathrm{adj}}(\Delta Z) = d$$

where $W = RR^T$

• eliminate ΔZ from (1):

$$\mathcal{A}^{\mathsf{adj}}(W^{-1}\mathcal{A}(\Delta y)W^{-1}) = d + \mathcal{A}^{\mathsf{adj}}(W^{-1}DW^{-1}) \tag{2}$$

a positive definite set of linear equations in Δy , and usually dense

total cost: cost of forming the equations (2) plus cost of solving

SDP with structure

minimize
$$q^T x$$

subject to $\begin{bmatrix} 0 & 0 \\ 0 & P \end{bmatrix} - \begin{bmatrix} P & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=1}^p x_i M_i \succeq N$

•
$$p + n(n+1)/2$$
 variables x , P

• we will assume that
$$p = O(n)$$

• discussion extends to problems with multiple constraints

$$\begin{bmatrix} 0 & 0 \\ 0 & P_k \end{bmatrix} - \begin{bmatrix} P_k & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=1}^p x_i M_{ki} \succeq N_k, \quad k = 1, \dots, L$$

Newton equations

$$W\Delta ZW + \begin{bmatrix} 0 & 0 \\ 0 & \Delta P \end{bmatrix} - \begin{bmatrix} \Delta P & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=1}^{p} \Delta x_i M_i = D_1$$
$$\Delta Z_{br} - \Delta Z_{tl} = D_2$$
$$\mathbf{Tr}(M_i \Delta Z) = d_i, \quad i = 1, \dots, p$$

where $\Delta Z_{\rm tl}$, $\Delta Z_{\rm br}$ are leading and trailing $n \times n$ submatrices of ΔZ

$$\Delta Z = \begin{bmatrix} \Delta Z_{\mathsf{tl}} \\ & \\ & \\ \end{bmatrix} = \begin{bmatrix} \Delta Z_{\mathsf{br}} \end{bmatrix}$$

 $W \succ 0$; value changes at each iteration

Standard solution method

$$W\Delta ZW + \begin{bmatrix} 0 & 0 \\ 0 & \Delta P \end{bmatrix} - \begin{bmatrix} \Delta P & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=1}^{p} \Delta x_i M_i = D_1$$
$$\Delta Z_{\mathsf{br}} - \Delta Z_{\mathsf{tl}} = D_2$$
$$\mathbf{Tr}(M_i \Delta Z) = d_i, \quad i = 1, \dots, p$$

- eliminate ΔZ from 1st equation
- solve dense set of equations in Δx , ΔP
- cost: cost of forming reduced equations plus at least $O(n^6)$ for solving

used in general-purpose solvers

Alternative method for solving Newton equations

first equation

$$W\Delta ZW + \begin{bmatrix} 0 & 0 \\ 0 & \Delta P \end{bmatrix} - \begin{bmatrix} \Delta P & 0 \\ 0 & 0 \end{bmatrix} + \sum_{i=1}^{p} \Delta x_i M_i = D_1$$

eliminate ΔP by taking inner product with E^k , $E = \begin{bmatrix} 0 & I_n \\ 0 & 0 \end{bmatrix}$:

$$\mathbf{Tr}(E^k W \Delta Z W) + \sum_{i=1}^p \Delta x_i \, \mathbf{Tr}(E^k M_i) = \mathbf{Tr}(E^k D_1), \quad k = 0, \dots, n$$

(note: $\mathbf{Tr}(E^kX)$ is sum of elements on kth diagonal of X)

second equation

$$\Delta Z_{\rm br} - \Delta Z_{\rm tl} = D_2$$

means

$$\Delta Z = \mathbf{T}(\Delta u) + Z_0$$
 for some Δu

• $\mathbf{T}(\Delta u)$ is symmetric Toeplitz matrix constructed from Δu

$$\mathbf{T}(\Delta u) = \begin{bmatrix} 2\Delta u_0 & \Delta u_1 & \cdots & \Delta u_n \\ \Delta u_1 & 2\Delta u_0 & \cdots & \Delta u_{n-1} \\ \vdots & \vdots & \ddots & \vdots \\ \Delta u_n & \Delta u_{n-1} & \cdots & 2\Delta u_0 \end{bmatrix}$$
$$= \sum_{k=0}^n \Delta u_k (E^k + (E^k)^T)$$

•
$$Z_0$$
 is any solution to $Z_{0,br} - Z_{0,tl} = D_2$

reduced Newton equations

$$\mathbf{Tr}(E^k W \mathbf{T}(\Delta u) W) + \sum_{i=1}^p \Delta x_i \mathbf{Tr}(E^k M_i) = \mathbf{Tr}(E^k D_1), \quad k = 0, \dots, n$$
$$\mathbf{Tr}(M_i \mathbf{T}(\Delta u)) = d_i, \quad i = 1, \dots, p$$

in matrix notation:

$$\begin{bmatrix} H & G \\ G^T & 0 \end{bmatrix} \begin{bmatrix} \Delta u \\ \Delta x \end{bmatrix} = \begin{bmatrix} r \\ d \end{bmatrix}$$

$$H_{ki} = \operatorname{Tr}\left(E^{k}W\left(E^{i} + (E^{i})^{T}\right)W\right), \qquad G_{ki} = \operatorname{Tr}(E^{k}M_{i})$$

total cost: $O(n^3)$ operations for solving, plus

- cost of forming G (can be pre-computed in at most $O(n^3)$, usually less)
- cost of forming H

fast evaluation of H via DFT

 $H_{ki} = \mathbf{Tr}(E^k W E^i W) + \mathbf{Tr}(E^k W (E^i)^T W), \quad i, k = 0, \dots, n$

• factor
$$W = RR^T = \sum_{l=0}^n r_l r_l^T$$

• take zero-padded (length $\geq 2(n+1)$) DFTs of r_l :

$$V = W_{\mathsf{DFT}}R$$

• evaluate *H* using Hadamard products:

$$H = W^H_{\mathsf{DFT}} \left((VV^H) \circ (VV^H)^T + (VV^T) \circ (VV^T)^H \right) W_{\mathsf{DFT}}$$
cost: $O(n^3)$

Summary

SDP formulation of a class of problems involving nonnegative polynomials:

- difficult to solve using general-purpose software
 - large number of auxiliary variables $(O(n^2))$
 - complexity typically $O(n^6)$ per iteration
- custom implementation of primal-dual interior-point method:
 - exploit structure in Newton equations using direct linear algebra
 - cost: $O(n^3)$ per iteration (times 20-30 iterations)

References

• Genin, Hachez, Nesterov, Van Dooren (SIMAX 2003)

nonnegative generalized matrix polynomials; fast implementations of dual barrier method using generalized Schur algorithm

Alkire, Vandenberghe (Mathematical Programming 2002)

trigonometric polynomials; fast implementation of dual barrier method using DFT

• Vandenberghe, Balakrishnan, Wallin, Hansson (this CDC)

extension to primal-dual methods, SDPs derived from KYP lemma