



Semidefinite programming

$$\begin{aligned} \min \quad & \langle C, X \rangle \\ \text{s.t.} \quad & A(X) = b \\ & X \succeq_{SD} 0 \end{aligned}$$

$$\begin{aligned} \mathcal{L}(X, \Lambda, V) &= \langle C, X \rangle - \langle \Lambda, X \rangle - \langle V, A(X) - b \rangle \\ &= \langle C - \Lambda - A^T(V), X \rangle + \langle b, V \rangle \\ &= \begin{cases} -\infty & C - \Lambda - A^T(V) \neq 0 \\ \langle b, V \rangle & C - \Lambda - A^T(V) = 0 \end{cases} \end{aligned}$$

$$\begin{aligned} \max_{\Lambda, V} \quad & \langle b, V \rangle \\ \text{s.t.} \quad & C - \Lambda - A^T(V) = 0 \\ & \Lambda \succeq_{SD} 0 \end{aligned}$$

Weak duality: if X is primal feasible and Λ, V are dual feasible

$$\langle C, X \rangle \geq \langle b, V \rangle$$

$$\langle C, X \rangle - \langle b, V \rangle = \langle C, X \rangle - \langle A(X), V \rangle = \langle C - A^T(V), X \rangle = \langle \Lambda, X \rangle \succeq 0$$

$$\begin{aligned} p^* = g(\Lambda^*, V^*) &= \inf_X \mathcal{L}(X, \Lambda^*, V^*) \\ &= \inf \langle C, X \rangle - \langle \Lambda^*, X \rangle - \langle A(X), V^* \rangle \\ &\leq \langle C, X^* \rangle - \langle \Lambda^*, X^* \rangle - \langle A(X^*), V^* \rangle \\ &\leq p^* \end{aligned}$$

$$\begin{aligned} \langle A(X), V \rangle &= \sum_i v_i \Lambda_i : X \\ &= \langle \sum_i v_i \Lambda_i, X \rangle \end{aligned}$$

eg. Strong duality of SDP may not hold

$$\begin{aligned} \min \quad & \gamma \\ \text{s.t.} \quad & \begin{bmatrix} 0 & \gamma \\ \gamma & 1 \end{bmatrix} \succeq_{SD} 0 \end{aligned} \xrightarrow{\text{primal}} \begin{aligned} \min_X \quad & \langle \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, X \rangle \\ \text{s.t.} \quad & \langle \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, X \rangle = 0 \\ & X \succeq_{SD} 0 \end{aligned}$$

primal is not strictly feasible

$$\begin{aligned} \max_{\Lambda, V} \quad & b \cdot V \\ \text{s.t.} \quad & \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} - \Lambda = V \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \\ & \Lambda \succeq_{SD} 0 \end{aligned}$$

$$X_{11} > 0 \implies 0/X_{11} - X_{11}^2 > 0$$

the primal optimal is 0 (when $\gamma_0 = 0$)

$$\begin{aligned} \left[\begin{array}{c} -V \ 1 \\ 1 \ 0 \end{array} \right] \succeq_{SD} 0 \\ -V \succeq 0 \quad -V \cdot 0 - 1 \succeq 0 \end{aligned} \quad \text{dual is infeasible}$$

$$\begin{aligned} \min \quad & \gamma \\ \text{s.t.} \quad & \begin{bmatrix} \gamma & 1 \\ 1 & \gamma \end{bmatrix} \succeq_{SD} 0 \end{aligned} \xrightarrow{\text{primal}} \begin{aligned} \min \quad & \langle \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, X \rangle \\ \text{s.t.} \quad & \langle \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, X \rangle = 2 \\ & X \succeq_{SD} 0 \end{aligned}$$

$$\begin{aligned} \gamma_0 > 0 \quad \gamma_1 > 0 \\ \gamma_1/\gamma_0 > 1 \end{aligned}$$

the optimal value is 0 but not attained

$$\begin{aligned} \max \quad & \gamma \cdot V \\ \text{s.t.} \quad & \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 & V \\ 0 & 0 \end{bmatrix} = \Lambda \\ & \Lambda \succeq 0 \end{aligned}$$

$$\begin{aligned} 0 - V^2 &\succeq 0 \\ \text{only if } V &\succeq 0 \end{aligned}$$

the dual is not strictly feasible
primal optimal is 0, but not attained

optimality condition

primal feasible: $X \succeq_{\text{H}} 0$ $A(X) = b$

dual feasible: $\Lambda \succeq_{\text{H}} 0$

complementary slackness: $X\Lambda = 0$

gradient variables: $A^T(U) + \Lambda = C$

$$\langle \Lambda, X \rangle = 0$$

$$\Lambda = PP^T \quad X = QQ^T$$

$$0 = \text{tr}(\Lambda X) = \text{tr}(PP^T QQ^T) = \text{tr}(P^T Q Q^T P) = \langle P^T Q, P^T Q \rangle$$

$$\therefore P^T Q = 0$$

$$\therefore \Lambda X = PP^T QQ^T = P0Q^T = 0$$

Barrier of Primal and Dual

$$\min_X \langle C, X \rangle$$

$$\text{s.t. } A(X) = b$$

$$X \succeq_{\text{H}} 0$$

↓

$$\min_X \langle C, X \rangle - \frac{1}{\epsilon} \log \det(X)$$

$$\text{s.t. } A(X) = b$$

$$\max_{\Lambda, U} \langle b, U \rangle$$

$$\text{s.t. } A^T(U) + \Lambda = C$$

$$\Lambda \succeq_{\text{H}} 0$$

↓

$$\min_{\Lambda} \langle b, U \rangle - \frac{1}{\epsilon} \log \det(\Lambda)$$

$$\text{s.t. } A^T(U) + \Lambda = C$$

$$\min_X \langle C, X \rangle - \frac{1}{\epsilon} \log \det X$$

$$\text{s.t. } A(X) = b$$

} primal barrier

$$\mathcal{L}(X, U) = \langle C, X \rangle - \frac{1}{\epsilon} \log \det(X) - \langle U, A(X) - b \rangle$$

$$= \langle C - A^T(U), X \rangle - \frac{1}{\epsilon} \log \det(X) + \langle U, b \rangle$$

$$\nabla_X \mathcal{L} = C - \frac{1}{\epsilon} X^{-1} - A^T(U) = 0 \quad X = \frac{1}{\epsilon} (C - A^T(U))^{-1}$$

$$g(U) = \inf_X \mathcal{L}(X, U)$$

$$= \begin{cases} -\infty & \text{if } C - A^T(U) \not\succeq_{\text{H}} 0 \\ \frac{1}{\epsilon} \log \det \left(\frac{1}{\epsilon} (C - A^T(U))^{-1} \right) + \langle U, b \rangle & \text{otherwise} \end{cases}$$

if $C - A^T(U)$ has a negative eigenvalue λ_i , with eigenvector u

$$\text{let } X = \alpha \cdot uu^T$$

$$\mathcal{L}(X, U) = \langle C - A^T(U), \alpha uu^T \rangle - \frac{1}{\epsilon} \log \det \alpha uu^T + \langle U, b \rangle \rightarrow -\infty \text{ as } \alpha \rightarrow +\infty$$

$$\max_{\Lambda, U} \frac{1}{\epsilon} \log \det \left(\frac{1}{\epsilon} (C - A^T(U))^{-1} \right) + \langle U, b \rangle$$

$$\text{s.t. } C - A^T(U) \succeq_{\text{H}} 0$$

⇒

$$\max_{\Lambda, U} \frac{1}{\epsilon} \log \det(\Lambda) + \langle U, b \rangle$$

$$\text{s.t. } C - A^T(U) = \Lambda$$

$$\Lambda \succeq_{\text{H}} 0$$

} dual barrier

KKT for primal barrier

primal feasible: $A(X) = b$ $X \succeq_{\text{H}} 0$

dual feasible: $\Lambda + A^T(U) = C$ $\Lambda \succeq_{\text{H}} 0$

gradient variables: $C - A^T(U) = \frac{1}{\epsilon} X^{-1}$

$$\Rightarrow X\Lambda = \frac{1}{\epsilon} I$$

$$\begin{array}{ll} \min & -\langle b, v \rangle - \frac{1}{\epsilon} \log \det(A) \\ \text{s.t.} & A^T(v) + \lambda = C \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{dual barrier}$$

$$\begin{aligned} \mathcal{L}(\lambda, v, x) &= -\langle b, v \rangle - \frac{1}{\epsilon} \log \det(A) + \langle x, A^T(v) + \lambda - C \rangle \\ &= -\langle A(v) - b, v \rangle - \frac{1}{\epsilon} \log \det(A) + \langle \lambda, x \rangle - \langle C, x \rangle \end{aligned}$$

$$\nabla_{\lambda} \mathcal{L} = -\frac{1}{\epsilon} \lambda^{-1} + x = 0 \quad \lambda^{-1} = \frac{1}{\epsilon} x$$

$$\begin{aligned} g(x) &= \inf_{\lambda, v} \mathcal{L} \\ &= \begin{cases} -\langle b, v \rangle & A(v) - b = 0 \\ -\infty & x \not\succeq 0 \\ -\frac{1}{\epsilon} \log \det(A) + \eta \frac{1}{\epsilon} - \langle C, x \rangle & \end{cases} \end{aligned}$$

$$\begin{array}{ll} \max_x & -\frac{1}{\epsilon} \log \det\left(\frac{1}{\epsilon} x^{-1}\right) + \eta \frac{1}{\epsilon} - \langle C, x \rangle \\ \text{s.t.} & A(x) = b \\ & x \succ_{\text{str}} 0 \end{array}$$

$$\Rightarrow \begin{array}{ll} \min_x & \langle C, x \rangle - \frac{1}{\epsilon} \log \det(x) \\ \text{s.t.} & A(x) = b \\ & x \succ_{\text{str}} 0 \end{array} \quad \left. \vphantom{\begin{array}{l} \min_x \\ \text{s.t.} \end{array}} \right\} \text{primal barrier}$$

primal barrier and dual barrier have the same optimality condition

if $x(\epsilon), \lambda(\epsilon), v(\epsilon)$ solve the optimality condition, then $x(\epsilon)$ solves the primal barrier and $\lambda(\epsilon), v(\epsilon)$ solves the dual barrier

KKT for dual barrier:

$$\begin{array}{ll} \text{primal feasible:} & \lambda \succ_{\text{str}} 0 \quad A^T(v) + \lambda = C \\ \text{dual feasible:} & x \succ_{\text{str}} 0 \quad A(x) = b \\ \text{gradient vanishes:} & \lambda = \frac{1}{\epsilon} x \end{array} \quad \left. \vphantom{\begin{array}{l} \text{primal feasible:} \\ \text{dual feasible:} \\ \text{gradient vanishes:} \end{array}} \right\}$$

Path-following Interior Point Method

generate $(x^k, \lambda^k, v^k) \approx (x(t^k), \lambda(t^k), v(t^k))$

key details $\left\{ \begin{array}{l} \text{proximity to central path} \\ \text{update: newton-like step} \end{array} \right.$

Local Neighborhoods of the Central Path

let $\mathcal{F}^0 = \{ (x, \lambda, v) : A(x) = b, A^T(v) + \lambda = C, x \succ 0, \lambda \succ 0 \}$ be the strictly feasible set

given $x \succ 0, \lambda \succ 0$ $u(x, \lambda) = \langle x, \lambda \rangle / n$

if x is primal feasible and λ, v is dual feasible $\langle C, x \rangle - \langle b, v \rangle = \langle C, x \rangle - \langle A(x), v \rangle = \langle x, \lambda \rangle$

$$d_{\mathcal{F}}(x, \lambda) = \| \lambda(x\lambda) - u(x, \lambda) \cdot \mathbf{I} \|_F$$

$$= \| x^{\frac{1}{2}} \lambda x^{\frac{1}{2}} - u(x, \lambda) \mathbf{I} \|_F$$

$$= \| \lambda^{\frac{1}{2}} x \lambda^{\frac{1}{2}} - u(x, \lambda) \mathbf{I} \|_F \quad (\text{measure of proximity of } x\lambda \text{ and } u(x, \lambda) \mathbf{I})$$

$$\mathcal{N}_{\mathcal{F}}(b) = \{ (x, \lambda, v) \in \mathcal{F}^0 : d_{\mathcal{F}}(x, \lambda) \leq \theta u(x, \lambda) \}$$

Newton Step

$$\begin{bmatrix} A^T(u) + \lambda - c \\ Ax - b \\ \lambda - \frac{1}{2}I \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \frac{1}{2}I - X\lambda \end{bmatrix} \quad X, \lambda \succ 0$$

↓ natural newton step

$$\begin{bmatrix} 0 & A^T & I \\ A & 0 & 0 \\ \lambda & 0 & X \end{bmatrix} \begin{bmatrix} \Delta X \\ \Delta v \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \frac{1}{2}I - X\lambda \end{bmatrix}$$

ΔX and $\Delta \lambda$ may not be symmetric
($\frac{1}{2}I - X\lambda$ may not be symmetric)

Search Direction

$$r_z(X, \lambda, v) = \begin{bmatrix} A^T(u) + \lambda - c \\ Ax - b \\ \lambda - \frac{1}{2}I \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} \langle \nabla f, A^T \rangle + \lambda - c \end{bmatrix} \\ \langle A, X \rangle - b \\ \langle X, X \rangle - \frac{1}{2}I \end{bmatrix} \begin{matrix} \text{rd dual residual} \\ \text{rp primal residual} \\ \text{rc centrality residual} \end{matrix}$$

$$r_z(X + \Delta X, \lambda + \Delta \lambda, v + \Delta v) \approx r_z(X, \lambda, v) + D r_z \cdot \begin{bmatrix} \Delta X \\ \Delta \lambda \\ \Delta v \end{bmatrix} := 0$$

$$\begin{matrix} \textcircled{1} \\ \textcircled{2} \\ \textcircled{3} \end{matrix} \begin{bmatrix} 0 & A^T(u) & I \\ A(u) & 0 & 0 \\ \lambda(u) & 0 & f(u) \end{bmatrix} \begin{bmatrix} \Delta X \\ \Delta v \\ \Delta \lambda \end{bmatrix} = -r_z(X, \lambda, v) = - \begin{bmatrix} \text{rd} \\ \text{rp} \\ \text{rc} \end{bmatrix}$$

$$\begin{aligned} \Delta X &= \Sigma^{-1}(-r_c - f(\Delta \lambda)) \quad \text{from (3)} \\ &= \Sigma^{-1} \begin{bmatrix} -r_c - f(-rd - A^T(\Delta v)) \end{bmatrix} \quad \text{from (1)} \end{aligned}$$

$$A \left(\Sigma^{-1} \begin{bmatrix} -r_c - f(-rd - A^T(\Delta v)) \end{bmatrix} \right) = -r_p \quad \text{from (2)}$$

$$A \Sigma^{-1}(-r_c - f(-rd)) + A \Sigma^{-1} f A^T(\Delta v) = -r_p$$

$$\Delta v = (A \Sigma^{-1} f A^T)^{-1} \begin{bmatrix} -r_p + A \Sigma^{-1}(r_c - f(-rd)) \end{bmatrix} \quad \text{solve for } \Delta v$$

$$\Delta \lambda = -rd - A^T(\Delta v) \quad \text{solve for } \Delta \lambda$$

$$\Delta X = \Sigma^{-1}(-r_c - f(\Delta \lambda)) \quad \text{solve for } \Delta X$$

Path Following Method

$$\begin{array}{ll} \min. <C, x> \\ \text{s.t.} & A(x) = b \\ & x \geq 0 \end{array} \quad \longrightarrow \quad \begin{array}{ll} \min & t <C, x> - \log \det(x) \\ \text{s.t.} & A(x) = b \end{array}$$

$$\tilde{L}(\tilde{x}, \tilde{v}) = t <C, \tilde{x}> - \log \det(\tilde{x}) - \langle \tilde{v}, A(\tilde{x}) - b \rangle$$

$$\nabla_{\tilde{x}} \tilde{L} = tC - \tilde{x}^{-1} - A^T(\tilde{v}) = 0$$

$$C - \frac{1}{t} \tilde{x}^{-1} - A^T(\tilde{v}/t) = 0 \quad \Rightarrow \quad \nabla_{\tilde{x}} \tilde{L} = \frac{d}{dt} <C, x> - \langle \lambda, x \rangle - \langle \tilde{v}/t, A(x) - b \rangle$$

if $\tilde{x}, \tilde{v}$ solves the hy-barrier, then $\tilde{x}$ minimize the Lagrangian

$$\mathcal{L}(x, \lambda, v) = <C, x> - \langle \lambda, x \rangle - \langle v, A(x) - b \rangle$$

$$\text{where } \lambda = \frac{1}{t} \tilde{x}^{-1}, \quad v = \tilde{v}/t$$

$$\begin{aligned} p^* &\geq g(\lambda, v) = \inf_x \mathcal{L}(x, \lambda, v) \\ &= <C, \tilde{x}> - \langle \lambda, \tilde{x} \rangle - \langle v, A(\tilde{x}) - b \rangle \\ &= <C, \tilde{x}> - \frac{1}{t} \langle \tilde{x}^{-1}, \tilde{x} \rangle \\ &= <C, \tilde{x}> - \frac{n}{t} \quad \text{duality gap} \end{aligned}$$

when x is primal feasible, λ, v is dual feasible

$$\langle \lambda, x \rangle = n/t \text{ is the duality gap}$$

$$t = n / \langle \lambda, x \rangle$$

while the:

$$\text{if } A^T(v) + \lambda - C = 0 \text{ and } A(x) = b \text{ and } \langle \lambda, x \rangle \leq \epsilon$$

return x, λ, v

$$t \leftarrow n / \langle \lambda, x \rangle \cdot \alpha \quad // \alpha \text{ controls the advancement of central path}$$

solve the system (Solve the hy-barrier)

$$\begin{cases} A^T(v + \alpha v) + (\lambda + \alpha \lambda) - C = 0 \\ A(x + \alpha x) = b \\ (x + \alpha x)(\lambda + \alpha \lambda) = \frac{1}{t} I \end{cases}$$

line search to get s

$$x \leftarrow x + s \cdot \alpha x$$

$$\lambda \leftarrow \lambda + s \cdot \alpha \lambda$$

$$v \leftarrow v + s \cdot \alpha v$$