

Quadratic Programming

Equality Constrained QP

John Bagterp Jørgensen

Department of Applied Mathematics and Computer Science
Technical University of Denmark
jbjo@dtu.dk

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Convex Equality Constrained Quadratic Program

Convex equality constrained quadratic program

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (1a)$$

$$s.t. \quad a_i'x = b_i \quad i \in \mathcal{E} = \{1, 2, \dots, m\} \quad (1b)$$

$H \in \mathbb{R}^{n \times n}$ is symmetric and positive-definite. $g \in \mathbb{R}^n$. $m < n$.

$A = [a_1 \ a_2 \ \dots \ a_m]$ has full column rank.

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (2a)$$

$$s.t. \quad A'x = b \quad (2b)$$

Optimality Conditions

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (3a)$$

$$s.t. \quad a'_i x = b_i \quad i \in \mathcal{E} = \{1, 2, \dots, m\} \quad (3b)$$

Lagrangian

$$\mathcal{L} = \frac{1}{2}x'Hx + g'x - \sum_{i \in \mathcal{E}} \lambda_i (a'_i x - b_i) \quad (4)$$

$$\nabla_x \mathcal{L} = Hx + g - \sum_{i \in \mathcal{E}} a_i \lambda_i \quad (5)$$

Optimality Conditions

$$Hx + g - \sum_{i \in \mathcal{E}} a_i \lambda_i = 0 \quad (6a)$$

$$a'_i x - b_i = 0 \quad i \in \mathcal{E} \quad (6b)$$

Optimality Conditions

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (7a)$$

$$s.t. \quad A'x = b \quad (7b)$$

Lagrangian

$$\mathcal{L} = \frac{1}{2}x'Hx + g'x - \lambda'(Ax - b) \quad (8)$$

$$\nabla_x \mathcal{L} = Hx + g - A\lambda \quad (9)$$

Optimality Conditions

$$Hx + g - A\lambda = 0 \quad (10a)$$

$$A'x - b = 0 \quad (10b)$$

KKT System

$$\begin{bmatrix} H & -A' \\ -A' & 0 \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} = - \begin{bmatrix} g \\ b \end{bmatrix} \quad (11)$$

Convex Equality Constrained QP & KKT System

Convex equality constrained quadratic program

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (12a)$$

$$s.t. \quad a'_i x = b_i \quad i \in \mathcal{E} = \{1, 2, \dots, m\} \quad (12b)$$

$H \in \mathbb{R}^{n \times n}$ is symmetric and positive-definite. $g \in \mathbb{R}^n$. $m < n$.

$A = [a_1 \ a_2 \ \dots \ a_m]$ has full column rank.

$$\min_{x \in \mathbb{R}^n} \quad \phi = \frac{1}{2}x'Hx + g'x \quad (13a)$$

$$s.t. \quad A'x = b \quad (13b)$$

The convex equality constrained QP is solved by solution of the KKT system

$$\begin{bmatrix} H & -A \\ -A' & 0 \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} = - \begin{bmatrix} g \\ b \end{bmatrix} \quad (14)$$

Direct Methods

$$\overbrace{\begin{bmatrix} H & -A \\ -A' & 0 \end{bmatrix}}^{=K} \overbrace{\begin{bmatrix} x \\ \lambda \end{bmatrix}}^{=z} = - \overbrace{\begin{bmatrix} g \\ b \end{bmatrix}}^{=d} \quad (15)$$

$x \in \mathbb{R}^n$, $\lambda \in \mathbb{R}^m$, $g \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, $H \in \mathbb{R}^{n \times n}$, $A \in \mathbb{R}^{n \times m}$

Dense Methods – The KKT matrix is represented as a dense matrix

Sparse Methods – The KKT matrix is represented as a sparse matrix

- ▶ backslash: $z = K \setminus d$;
- ▶ LU factorization: $PK = LU$
 1. `[L,U,p] = lu(K,'vector');`
 2. $z = U \setminus (L \setminus d(p))$;
- ▶ LDL factorization: $P'KP = LDL'$
 1. `z = zeros(n+m,1);`
 2. `[L,D,p] = ldl(K,'lower','vector');`
 3. $z(p) = L' \setminus (D \setminus (L \setminus d(p)))$;

Range-Space Method / Schur-Complement Method

$$\begin{bmatrix} H & -A \\ -A' & 0 \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} = - \begin{bmatrix} g \\ b \end{bmatrix} \quad H \in \mathbb{S}_{++}^{n \times n} \quad (16)$$

$$Hx - A\lambda = -g \quad \Leftrightarrow \quad x = H^{-1}A\lambda - H^{-1}g \quad (17)$$

$$b = A'x = (A'H^{-1}A)\lambda - A'H^{-1}g \quad (18)$$

Procedure

1. Cholesky factorize $H = LL'$.
2. Solve $Hv = g$ for v .
3. Form $H_A = A'H^{-1}A = L_AL'_A$ and its factorization (do not form H^{-1}).
4. Solve $H_A\lambda = b + A'v$ for λ .
5. Solve $Hx = A\lambda - g$ for x .

Useful when

- ▶ H is well-conditioned and easy to invert (H is diagonal or block-diagonal)
- ▶ H^{-1} is known explicitly, e.g. through quasi-Newton update formulas
- ▶ The number of equality constraints (m) is small, i.e. $H_A = A'H^{-1}A \in \mathbb{R}^{m \times m}$ is small.

Null-Space Method

$$\begin{bmatrix} H & -A \\ -A' & 0 \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} = - \begin{bmatrix} g \\ b \end{bmatrix} \quad A \in \mathbb{R}^{n \times m} \quad (19)$$

Define the non-singular matrix $\begin{bmatrix} Y & Z \end{bmatrix} \in \mathbb{R}^{n \times n}$ with $Y \in \mathbb{R}^{n \times m}$ and $Z \in \mathbb{R}^{n \times (n-m)}$ such that

$$A' \begin{bmatrix} Y & Z \end{bmatrix} = \begin{bmatrix} A'Y & A'Z \end{bmatrix} = \begin{bmatrix} A'Y & 0 \end{bmatrix} \quad A'Y \in \mathbb{R}^{m \times m} \text{ non-singular}$$

$$x = Yx_Y + Zx_Z$$

$$b = A'x = A'Yx_Y + A'Zx_Z = A'Yx_Y$$

$$-g = Hx - A\lambda = HYx_Y + HZx_Z - A\lambda$$

$$(Z'HZ)x_Z = -Z'(HYx_Y + g)$$

$$(A'Y)' \lambda = Y'(Hx + g)$$

Procedure

1. Solve: $(A'Y)x_Y = b$
2. Solve: $(Z'HZ)x_Z = -Z'(HYx_Y + g)$
3. Compute: $x = Yx_Y + Zx_Z$
4. Solve: $(A'Y)' \lambda = Y'(Hx + g)$

Properties

- Useful when the number of degrees of freedom, $n - m$, is small.
- Main drawback is its need for the null-space matrix, Z , which may be expensive to compute.

Null-Space Method: Orthonormal basis

$$A = Q \begin{bmatrix} R \\ 0 \end{bmatrix} = [Q_1 \quad Q_2] \begin{bmatrix} R \\ 0 \end{bmatrix} = Q_1 R \quad (20)$$

$$A' = [R' \quad 0] Q' \quad (21)$$

$$A'Q = A' [Q_1 \quad Q_2] = [A'Q_1 \quad A'Q_2] = [R' \quad 0] \quad (22)$$

$$\begin{bmatrix} Y & Z \end{bmatrix} = [Q_1 \quad Q_2] \quad A'Y = A'Q_1 = R' \quad (23)$$

QR factorization in Matlab ($A \in \mathbb{R}^{n \times m}$)

1. `[Q,Rbar] = qr(A);`
2. `m1 = size(Rbar,2);`
3. `Q1 = Q(:,1:m1); Q2 = Q(:,m1+1:n); R = Rbar(1:m1,1:m1)`

If $A \in \mathbb{R}^{n \times m}$ has full column rank $m_1 = m$.

Procedure

1. Solve: $R'x_Y = b$
2. Solve: $(Q_2' H Q_2)x_Z = -Q_2'(H Q_1 x_Y + g)$
3. Compute: $x = Q_1 x_Y + Q_2 x_Z$
4. Solve: $R\lambda = Q_1'(Hx + g)$