

# KKT Matrix Conditioning

## Convex Optimization

Major sources:  
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# Quadratic Optimization & KKT System

$$\min_x \quad \frac{1}{2}x^\top Qx + c^\top x$$

$$\text{s.t.} \quad Ax = b$$

Primal Variables:  $x \in \mathbb{R}^n$

Dual Variables:  $v \in \mathbb{R}^m$

$$Q \in \mathbb{R}^{n \times n} \quad Q = Q^\top \succ 0 \quad c \in \mathbb{R}^n$$

$$A \in \mathbb{R}^{m \times n} \quad b \in \mathbb{R}^m$$

fat ( $m < n$ ),

full row rank ( $\text{rank} = m$ )

**Lagrangian:**  $\mathcal{L}(x, v) = \frac{1}{2}x^\top Qx + c^\top x + v^\top (Ax - b)$

**KKT Conditions:**

1. Stationarity:  $x^\top Q + c^\top + v^\top A = 0$

2. Feasibility:  $Ax - b = 0$

**KKT System:**

$$\begin{bmatrix} Q & A^\top \\ A & 0 \end{bmatrix} \begin{bmatrix} x \\ v \end{bmatrix} = \begin{bmatrix} -c \\ b \end{bmatrix}$$

KKT Matrix

**Solution:**

$$\begin{bmatrix} x \\ v \end{bmatrix} = \begin{bmatrix} Q & A^\top \\ A & 0 \end{bmatrix}^{-1} \begin{bmatrix} -c \\ b \end{bmatrix}$$

# KKT Solutions

$$\begin{array}{ll} \min_x & \frac{1}{2}x^\top Qx + c^\top x \\ \text{s.t.} & Ax = b \end{array}$$

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KKT Matrix

**Solution:**

$$\begin{bmatrix} x \\ v \end{bmatrix} = \begin{bmatrix} Q^{-1} - Q^{-1}A^\top(AQ^{-1}A^\top)^{-1}AQ^{-1} & Q^{-1}A^\top(AQ^{-1}A^\top)^{-1} \\ (AQ^{-1}A^\top)^{-1}AQ^{-1} & -(AQ^{-1}A^\top)^{-1} \end{bmatrix} \begin{bmatrix} -c \\ b \end{bmatrix}$$

...using **block matrix inversion** (or just directly verify).

# KKT Solutions

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KKT Matrix

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*...projection matrix*

*...using **block matrix inversion** (or just directly verify).*

# KKT Solutions

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KKT Matrix

## Solution:

$$x = Q^{-1}A^\top (AQ^{-1}A^\top)^{-1} (AQ^{-1}c + b) - Q^{-1}c$$

$$v = -(AQ^{-1}A^\top)^{-1} (AQ^{-1}c + b)$$

# KKT Matrix Properties

$$\begin{array}{ll} \min_x & \frac{1}{2}x^\top Qx + c^\top x \\ \text{s.t.} & Ax = b \end{array}$$

Primal Variables:  $x \in \mathbb{R}^n$

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KKT Matrix

## Matrix Properties:

$$M = \begin{bmatrix} Q & A^\top \\ A & 0 \end{bmatrix}$$

$M$  indefinite

$$\text{spec}(M) = \rho_+ \sqcup \rho_-$$

**Positive:**  $\rho_+ = \{\lambda_1, \dots, \lambda_n\} > 0$

**Negative:**  $\rho_- = \{\lambda_{n+1}, \dots, \lambda_{n+m}\} < 0$

## Cardinality

$$|\rho_+| = n$$

$$|\rho_-| = m$$

# Conditioned KKT Matrix: Option 1\*

\*many conditionings are possible

## Coordinate Transforms

$$\begin{aligned} \min_x \quad & \frac{1}{2}x'^{\top} P^{\top} Q P x' + c^{\top} P x' \\ \text{s.t.} \quad & W^{\top} A P x' = W^{\top} b \end{aligned}$$

$$\begin{aligned} \text{Primal: } \quad & x = P x' \in \mathbb{R}^n \\ \text{Dual: } \quad & v = W v' \in \mathbb{R}^m \end{aligned}$$

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### KKT Conditions:

- Stationarity:  $x'^{\top} P^{\top} Q P + c^{\top} P + v'^{\top} W^{\top} A P = 0$
- Feasibility:  $W^{\top} A P x' - W^{\top} b = 0$

### KKT System:

$$\begin{bmatrix} P^{\top} Q P & P^{\top} A^{\top} W \\ W^{\top} A P & 0 \end{bmatrix} \begin{bmatrix} x' \\ v' \end{bmatrix} = \begin{bmatrix} -P^{\top} c \\ W^{\top} b \end{bmatrix}$$

### Conditioning:

$$M \Rightarrow \begin{bmatrix} P^{\top} Q P & P^{\top} A^{\top} W \\ W^{\top} A P & 0 \end{bmatrix} = \begin{bmatrix} P^{\top} & 0 \\ 0 & W^{\top} \end{bmatrix} \begin{bmatrix} Q & A^{\top} \\ A & 0 \end{bmatrix} \begin{bmatrix} P & 0 \\ 0 & W \end{bmatrix}$$

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**Conditioning:**  $A Q^{-\frac{1}{2}} = U [\Sigma \ 0] V^{\top}$  (SVD)  $\Rightarrow$  Take  $P = Q^{-\frac{1}{2}} V$   $W = U$

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$$M \Rightarrow \begin{bmatrix} P^{\top} Q P & P^{\top} A^{\top} W \\ W^{\top} A P & 0 \end{bmatrix} = \begin{bmatrix} I & 0 & \Sigma \\ 0 & I & 0 \\ \Sigma & 0 & 0 \end{bmatrix} \xrightarrow{\text{Permute...}} \begin{bmatrix} I & \Sigma & 0 \\ \Sigma & 0 & 0 \\ 0 & 0 & I \end{bmatrix}$$

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**Subblocks:**

$$\begin{bmatrix} 1 & \sigma_j \\ \sigma_j & 0 \end{bmatrix} \quad \begin{aligned} s(s-1) - \sigma_j^2 &= 0 \\ \lambda_{j,j+1} &= \frac{1}{2} \pm \sqrt{\frac{1}{4} + \sigma_j^2} \end{aligned}$$

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## Coordinate Transforms

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## KKT System:

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**Conditioning:**  $A Q^{-\frac{1}{2}} = U [\Sigma \ 0] V^{\top}$  (SVD)  $\Rightarrow$  Take  $P = Q^{-\frac{1}{2}} V$   $W = U$

$$M \Rightarrow \begin{bmatrix} P^{\top} Q P & P^{\top} A^{\top} W \\ W^{\top} A P & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} I & \Sigma & 0 \\ \Sigma & 0 & 0 \\ 0 & 0 & I \end{bmatrix}$$

## Subblocks:

$$\begin{bmatrix} 1 & \sigma_j \\ \sigma_j & 0 \end{bmatrix}$$

$$\begin{aligned} & s(s-1) - \sigma_j^2 = 0 & \lambda_{j,j+1} &= \frac{1}{2} \pm \sqrt{\frac{1}{4} + \sigma_j^2} \\ & \times (n-m) & \times m & \times m \\ & \downarrow & \downarrow & \downarrow \\ \text{spec} &= \underbrace{\left\{ 1, \frac{1}{2} + \sqrt{\frac{1}{4} + \sigma_j^2}, \frac{1}{2} - \sqrt{\frac{1}{4} + \sigma_j^2} \right\}}_{+} \end{aligned}$$

# Conditioned KKT Matrix: Option 2\*

\*many conditionings are possible

## Coordinate Transforms

$$\begin{aligned} \min_x \quad & \frac{1}{2} x'^{\top} P^{\top} Q P x' + c^{\top} P x' \\ \text{s.t.} \quad & W^{\top} A P x' = W^{\top} b \end{aligned}$$

$$\begin{aligned} \text{Primal: } \quad & x = P x' \in \mathbb{R}^n \\ \text{Dual: } \quad & v = W v' \in \mathbb{R}^m \end{aligned}$$

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**Lagrangian:**  $\mathcal{L}(x', v') = \frac{1}{2} x'^{\top} P^{\top} Q P x' + c^{\top} P x' + v'^{\top} W^{\top} (A P x' - b)$

### KKT Conditions:

- Stationarity:  $x'^{\top} P^{\top} Q P + c^{\top} P + v'^{\top} W^{\top} A P = 0$
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$$M \Rightarrow \begin{bmatrix} P^{\top} Q P & P^{\top} A^{\top} W \\ W^{\top} A P & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} I & I & 0 \\ I & 0 & 0 \\ 0 & 0 & I \end{bmatrix}$$

### Subblocks:

$$\begin{aligned} & s(s-1) - 1 = 0 & \lambda_{j,j+1} &= \frac{1}{2} \pm \sqrt{\frac{1}{4} + 1} \\ & \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} & \begin{matrix} \times (n-m) & \times m & \times m \\ \downarrow & \downarrow & \downarrow \end{matrix} & \\ \text{spec} &= \underbrace{\left\{ 1, \frac{1}{2} + \sqrt{\frac{1}{4} + 1} \right\}}_{+} \underbrace{\left\{ \frac{1}{2} - \sqrt{\frac{1}{4} + 1} \right\}}_{-} \end{aligned}$$

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### Subblocks:

$$\begin{aligned} & s(s-1) - 1 = 0 \quad \lambda_{j,j+1} = \frac{1 \pm \sqrt{5}}{2} \\ & \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \quad \begin{matrix} \times (n-m) & \times m & \times m \\ \downarrow & \downarrow & \downarrow \end{matrix} \\ \text{spec} = & \left\{ \underbrace{1, \frac{1+\sqrt{5}}{2}}_{+}, \underbrace{\frac{1-\sqrt{5}}{2}}_{-} \right\} \end{aligned}$$