

Semidefinite Relaxations of Quadratically Constrained Quadratic Programs

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This document provides a derivation of the semidefinite relaxation for quadratically constrained quadratic (QCQP) programs and shows it is the bidual (dual of the dual) of the QCQP.

1 Quadratically Constrained Quadratic Programs

A homogeneous QCQP is given by

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \mathbf{x}^\top \mathbf{Q} \mathbf{x} \\ & \text{subject to} && \mathbf{x}^\top \mathbf{A}_i \mathbf{x} + b_i = 0, \quad i = 1, \dots, m \end{aligned} \tag{QCQP}$$

The Lagrangian is given by

$$L(\mathbf{x}, \boldsymbol{\lambda}) = \mathbf{x}^\top \mathbf{Q} \mathbf{x} + \sum_{i=1}^m \lambda_i (b_i + \mathbf{x}^\top \mathbf{A}_i \mathbf{x}) \tag{1}$$

$$= \boldsymbol{\lambda}^\top \mathbf{b} + \mathbf{x}^\top \left(\mathbf{Q} + \sum_{i=1}^m \lambda_i \mathbf{A}_i \right) \mathbf{x}, \tag{2}$$

such that the dual problem is given by

$$\begin{aligned} & \underset{\boldsymbol{\lambda}}{\text{maximize}} && \boldsymbol{\lambda}^\top \mathbf{b} \\ & \text{subject to} && \mathbf{Q} + \sum_{i=1}^m \lambda_i \mathbf{A}_i \succcurlyeq 0 \end{aligned} \tag{Dual}$$

Notice that the PSD constraint is given by half of the Hessian of the Lagrangian (2) with respect to $\mathbf{x}$, which is denoted

$$\mathcal{H}(\boldsymbol{\lambda}) = \frac{1}{2} \frac{\partial^2 L(\boldsymbol{\lambda}, \mathbf{x})}{\partial \mathbf{x} \partial \mathbf{x}^\top} = \mathbf{Q} + \sum_{i=1}^m \lambda_i \mathbf{A}_i. \tag{3}$$

To derive the bidual, the Lagrangian of the dual problem is given by

$$L(\boldsymbol{\lambda}, \mathbf{S}) = \boldsymbol{\lambda}^\top \mathbf{b} + \left\langle \mathbf{Q} + \sum_{i=1}^m \lambda_i \mathbf{A}_i, \mathbf{S} \right\rangle \quad (4)$$

$$= \boldsymbol{\lambda}^\top \mathbf{b} + \text{tr} \left(\left(\mathbf{Q} + \sum_{i=1}^m \lambda_i \mathbf{A}_i \right) \mathbf{S} \right) \quad (5)$$

$$= \sum_{i=1}^m \lambda_i b_i + \text{tr}(\mathbf{Q}\mathbf{S}) + \sum_{i=1}^m \lambda_i \text{tr}(\mathbf{A}_i \mathbf{S}) \quad (6)$$

$$= \text{tr}(\mathbf{Q}\mathbf{S}) + \sum_{i=1}^m \lambda_i (b_i + \text{tr}(\mathbf{A}_i \mathbf{S})), \quad (7)$$

where $\mathbf{S} \succcurlyeq 0$ is the bidual Lagrange multiplier and $\boldsymbol{\lambda}$ are the Lagrange multipliers of the original dual. The inner product notation is used as $\langle \mathbf{A}, \mathbf{B} \rangle = \sum_{i=1}^M \sum_{j=1}^N \mathbf{A}_{ij} \mathbf{B}_{ij} = \text{tr}(\mathbf{A}\mathbf{B}^\top)$. Since this is the Lagrangian of a maximization problem, the constraint cost terms are added instead of subtracted. The bidual dual function is given by

$$g(\mathbf{S}) = \sup_{\boldsymbol{\lambda}} \left(\text{tr}(\mathbf{Q}\mathbf{S}) + \sum_{i=1}^m \lambda_i (b_i + \text{tr}(\mathbf{A}_i \mathbf{S})) \right), \quad (8)$$

which means that $b_i + \text{tr}(\mathbf{A}_i \mathbf{S}) = 0$ for feasibility of the bidual optimization problem, which is given by

$$\begin{aligned} & \underset{\mathbf{S}}{\text{minimize}} && \langle \mathbf{Q}, \mathbf{S} \rangle && \text{(Bidual-SDP)} \\ & \text{subject to} && \langle \mathbf{A}_i, \mathbf{S} \rangle + b_i = 0. \end{aligned}$$

The problems (Bidual-SDP) and (Dual) are the duals of each other. Furthermore, (Dual) is the dual of the non-convex QCQP (QCQP). The optimal values of the three problems are related by $\text{val}(\text{Dual}) \leq \text{val}(\text{Bidual-SDP}) \leq \text{val}(\text{QCQP})$.

Furthermore, (Bidual-SDP) is a rank relaxation of (QCQP). Since $\mathbf{x}^\top \mathbf{Q} \mathbf{x} = \text{tr}(\mathbf{x} \mathbf{x}^\top \mathbf{Q})$, letting $\mathbf{Z} = \mathbf{x} \mathbf{x}^\top$ yields an equivalent problem to (QCQP) as

$$\begin{aligned} & \underset{\mathbf{Z}}{\text{minimize}} && \langle \mathbf{Z}, \mathbf{Q} \rangle \\ & \text{subject to} && \langle \mathbf{Z}, \mathbf{Q} \rangle + b_i = 0, \quad i = 1, \dots, m \\ & && \text{rank}(\mathbf{Z}) = 1. \end{aligned}$$

Relaxing the rank constraint into a positive semidefinite constraint yields exactly (Bidual-SDP).