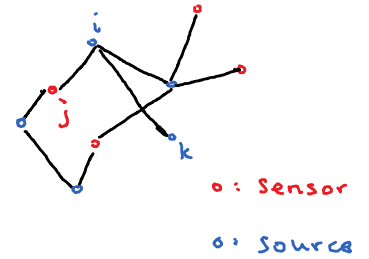


Example: Sensor network localization

$x_i \in \mathbb{R}^d$. Unknown position of i^{th} source

$a_j \in \mathbb{R}^d$: known position of j^{th} sensor



Assuming without measurement noise, we have

$$(L) \begin{cases} \|x_i - x_k\|_2^2 = d_{ik}^2, & \forall (i,k) \in E \quad \text{non-convex} \\ \|x_i - a_j\|_2^2 = \bar{d}_{ij}^2, & \forall (i,j) \in E \quad \text{non-convex} \end{cases}$$

edge set

Goal: Find $x_1, \dots, x_n \in \mathbb{R}^d$ that satisfy the above quadratic equations.

Apply SDR to (L):

$$\|x_i - x_k\|_2^2 = \underbrace{x_i^T x_i}_{\sum_{l=1}^d x_{il}^2} - 2 \underbrace{x_i^T x_k}_{\sum_{l=1}^d x_{il} x_{kl}} + x_k^T x_k$$

too tedious

This gives the substitution: $Y_{ik} = x_i^T x_k$. Hence, define

$$Y = \begin{matrix} & \begin{matrix} \text{---} \\ \text{---} \\ \text{---} \end{matrix} \\ \begin{matrix} \text{---} \\ \text{---} \\ \text{---} \end{matrix} \\ \begin{matrix} \text{---} \\ \text{---} \\ \text{---} \end{matrix} \end{matrix} X^T X, \quad \text{where } X = \begin{bmatrix} | & & | \\ x_1 & \dots & x_n \\ | & & | \end{bmatrix} \in \mathbb{R}^{d \times n}$$

(Compare $Y = XX^T$ in the previous approach)

Hence,

$$\|x_i - x_k\|_2^2 = Y_{ii} - 2Y_{ik} + Y_{kk} = \left\langle \underbrace{\begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}}_{E_{ik}} \begin{matrix} i \\ k \end{matrix}, Y \right\rangle$$

Moreover,

$$\|x_i - a_j\|_2^2 = \underbrace{x_i^T x_i}_{Y_{ii}} - 2 \underbrace{a_j^T x_i}_{\text{linear in } x_i} + \underbrace{a_j^T a_j}_{\text{constant}}$$

For any $A, B \in \mathbb{R}^{m \times n}$,

$$\langle A, B \rangle = \text{tr}(A^T B)$$

$$a_j^T x_i = a_j^T X e_i = \text{tr}(a_j^T X e_i)$$

$$= \left\langle \begin{bmatrix} a_j^T | & -a_j^T | \\ \hline -e_i^T | & E_i \end{bmatrix}, \begin{bmatrix} I & X \\ X^T & Y \end{bmatrix} \right\rangle$$

" $E_i = \begin{bmatrix} | & \\ 1 & \end{bmatrix} \begin{bmatrix} I \\ v^T \end{bmatrix} \begin{bmatrix} I & X \end{bmatrix}$

$$\begin{aligned} & \langle a_j^T, I \rangle \\ &= \text{tr}(a_j^T a_j) \\ &= \text{tr}(a_j^T a_j) \\ &= a_j^T a_j \end{aligned}$$

$$\begin{aligned} \gamma_j^2 &= a_j^T a_j \\ &= \text{tr}(a_j^T x e_i) \\ &= \text{tr}(x e_i a_j^T) \\ &= \langle a_j e_i^T, x \rangle \end{aligned}$$

$$E_i = \begin{bmatrix} i & \\ & 1 \end{bmatrix}; \begin{bmatrix} I \\ x^T \end{bmatrix} [I \ x] \quad \text{tr}(a_j a_j) = a_j^T a_j$$

Hence, (L) is equivalent to

$$\left\langle \begin{bmatrix} 0 & 0 \\ 0 & E_{ik} \end{bmatrix}, \begin{bmatrix} I & x \\ x^T & Y \end{bmatrix} \right\rangle = d_{ik}^2, \quad \leftarrow \text{linear}$$

$$\left\langle \begin{bmatrix} a_j^T & -a_j^T \\ -a_j^T & E_i \end{bmatrix}, \begin{bmatrix} I & x \\ x^T & Y \end{bmatrix} \right\rangle = d_{ij}^2, \quad \leftarrow \text{linear}$$

$$Y = X^T X \quad \leftarrow \text{non-convex} \quad \{ (x, Y) : Y = X^T X \}$$

The SDR approach is to relax $Y = X^T X$ to $Y - X^T X \in \mathcal{S}_+^n$.

By Schur complement,

$$Y - X^T X \in \mathcal{S}_+^n \iff \begin{bmatrix} I & X \\ X^T & Y \end{bmatrix} \in \mathcal{S}_+^{d+n} \iff \begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \succeq 0$$

$$\iff C - B^T A^{-1} B \succeq 0, \quad A, C \succeq 0$$

Example: Extension of SDR to complex QCQPs

Consider

$$\begin{aligned} \text{inf } z^H C z & \quad \text{always real: } (z^H C z)^H = z^H C^H z = z^H C z \\ \text{s.t. } z^H Q_i z & \geq b_i, \quad i=1, \dots, m. \\ z & \in \mathbb{C}^n \end{aligned}$$

Data:

$$C, Q_1, \dots, Q_m \in \mathbb{H}^n$$

$$b_1, \dots, b_m \in \mathbb{R}$$

set of $n \times n$ Hermitian matrices (i.e., $A = A^H$)

The SDR technique still applies:

$$z^H C z = \text{tr}(z^H C z) = \text{tr}(C z z^H) = \langle C, Z \rangle$$

This gives the following SDR of (C):

$$\text{inf } \langle C, Z \rangle$$

$$\text{s.t. } \langle Q_i, Z \rangle \geq b_i,$$

$$Z = z z^H$$

$$\iff z_{ij} = z_i \bar{z}_j$$

$$z \in \mathbb{H}_+^n$$

↖ set of $n \times n$ Hermitian
psd matrices; i.e.,

$$A \in \mathbb{H}_+^n \Leftrightarrow z^H A z \geq 0 \quad \forall z \in \mathbb{C}^n$$