## Global QCQP Solver

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## QCQP

Suppose we readily have a sparse pattern G(V, E) (and extension  $E \subseteq F$ ) for Q We denote aggregated pattern if G(V, E) stands for all data matrices.  $F = \bigcup_r C_r$ , where  $C_r$  the maximal cliques can be computed efficiently for chordal graph F. SD constraint  $Y \succeq 0$  is equivalent to,

$$Y_{(i,i)\in C_r} \succeq 0, \quad \forall r$$

Can be expressed as follows,

$$E_r \in \mathbf{R}^{|C_r| \times n}$$

$$(E_r)_{i,j} = \begin{cases} 1, & \text{if } C_r(i) = j \\ 0, & \text{otherwise} \end{cases}$$

$$Y_r \equiv E_r Y E_r^T \succeq 0$$

$$(1)$$

## Sparse-SDP: SSDP

In QCQP, we have the SD constraint:

$$Y - xx^T \succeq 0$$

then any subprinciple matrix is also semidefinite,

$$E_r(Y - xx^T)E_r^T \succeq 0 \tag{2}$$

So we have the QCQP with r smaller SDP constraints.

Maximize 
$$Q \bullet Y + q^T x$$
  
s.t.  $\begin{bmatrix} E_r Y E_r^T & E_r x \\ (E_r x)^T & 1 \end{bmatrix} \succeq 0, \forall r$  (3)

This formulation incorporates a lot of linear constraints for  $C_r \cap C_{r'}$ 

Consider the case for 2-blocks (2 cliques), let  $D_1 = C_1 \cap C_2$ 

$$Y = \sum_{r=1}^{2} E_r^T Y_r Er - E(D_1)^T Y_{D_1} E(D_1)$$
(4)

Maximize 
$$Q \bullet Y + q^T x$$
  
s.t.  $\begin{bmatrix} E_r Y E_r^T & E_r x \\ (E_r x)^T & 1 \end{bmatrix} \succeq 0, \forall r$  (5)

We see the above is equivalent to (3), if there is no intersections for blocks, i.e.,  $D_1 = \emptyset$ , it can be further simplified by letting  $Q_r = E_r Q E_r^T$ 

Maximize 
$$\sum_{r=1}^{L} Q_r \bullet Y_r$$
s.t. 
$$\begin{bmatrix} Y_r & E_r x \\ (E_r x)^T & 1 \end{bmatrix} \succeq 0, \forall r$$

$$\operatorname{diag}(Y_r) \leq E_r x$$
(6)

The same for any similarity transformations