- 1. SDP Relaxation and Heuristics for Two-Way Partitioning Problem
 - (a) Q 5.39 textbook

$$min \ x^T W x$$

$$s.t. \ x_i^2 = 1, \forall i \in \{1, .., n\}$$

i. Show that the two-way partitioning problem can be cast as

$$min \ tr(WX)$$

$$s.t. \ X \succeq 0, rank(X) = 1$$

$$X_{ii} = 1, \forall i \in \{1, ..., n\}$$

$$x^{T}Wx = tr(x^{T}Wx) = tr(Wxx^{T})$$

$$let \ X = xx^{T}$$

$$(\forall i)x_{i}^{2} = 1 \iff x_{i} = \{-1, 1\} \implies x^{T}Ix = n$$

$$x^{T}Ix = tr(xx^{T}) = n$$

$$((\exists i)X_{ii} = -1 \land (\forall i, j)X_{ij} = \{-1, 1\} \implies tr(X) < n)$$

$$thus, (\forall i)X_{ii} = 1 \text{ for } tr(X) = n$$

$$X = xx^{T} = x \begin{bmatrix} a_{1} & a_{2} & \dots & a_{n} \end{bmatrix} = \begin{bmatrix} a_{1}x & a_{2}x & \dots & a_{n}x \end{bmatrix}, a_{i} \in \mathbb{R}, x \in \mathbb{R}^{n}$$

$$(\forall i)(\exists j)\beta_{ij}a_{i}x = a_{j}x \implies \beta_{ij}a_{i}x - a_{j}x = 0$$

$$let \ \gamma_{ij} = \beta_{ij}a_{i} - a_{j}$$

$$\gamma_{ij}x = 0$$

 $x \neq 0 \implies (\forall i)(\exists j)\gamma_{ij} = 0 \implies linear dependence between column vectors of X thus, <math>rank(X) = 1$

$$(\forall w)w^T X w = w^T x x^T w = (x^T w)^T x^T w$$

$$(\forall i)(\forall w)(x^T w)_i (x^T w)_i \ge 0 \implies (\forall w)(x^T w)^T (x^T w) \ge 0 \iff X \text{ is } SPD$$

Combining all constraints and objective forms the desired result

ii. SDP relaxation of two-way partitioning problem. Using the formulation in part (a), we can form the relaxation:

$$mintr(WX)$$

$$s.t. \ X \succeq 0$$

$$X_{ii} = 1, \forall i \in \{1, ..., n\}$$

This problem is an SDP, and therefore can be solved efficiently. Explain why its optimal value gives a lower bound on the optimal value of the two-way partitioning problem (5.113). What can you say if an optimal point X^* for this SDP has rank one?

$$X \succeq 0 \land X_{ii} = 1, \forall i \in \{1, ..., n\} \implies (\forall i)\lambda_i(X) \ge 1$$

- (b) Q 11.23(b-d) textbook
- 2. Interior Point Method