

Eex- 58-a)

$$\max_{B_1, B_2, d} (B_1, B_2)^{1/2}$$

$$\text{s.t. } P_K^T B P_K - 2d^T P_K + d^T B^{-1} d \leq 1, \text{ for } k=1, \dots, K$$

$$B_1 > 0$$

$$B_2 > 0$$

where $B = \begin{bmatrix} B_1 & 0 \\ 0 & B_2 \end{bmatrix}$

Taking the first set of conditions:

$$P_K^T B P_K - 2d^T P_K + d^T B^{-1} d \leq 1 \Rightarrow (I - P_K^T B P_K + 2d^T P_K) - d^T B^{-1} d \geq 0$$

Schur's Complement

$$\Leftrightarrow \begin{bmatrix} B & d \\ d^T & \mu_k \end{bmatrix} \succ 0, \mu_k = 1 - P_K^T B P_K + 2d^T P_K$$

So the problem is equivalent to

$$\max_{B_1, B_2, d} (B_1, B_2)^{1/2}$$

subject to
variables

$$\text{s.t. } \begin{bmatrix} B & d \\ d^T & \mu_k \end{bmatrix} \succ 0, k=1, \dots, K$$

$$B > 0$$

$$\max_{B_1, B_2, d, \lambda} \lambda$$

$$\text{s.t. } \begin{bmatrix} B & d \\ d^T & \mu_k \end{bmatrix} \succ 0, k=1, \dots, K$$

$$B > 0$$

$$\lambda^2 = B_1, B_2$$

Relaxing the last condition to an inequality:

max λ

B_1, B_2, d, λ

$$\left[\begin{matrix} B_1 d \\ d^T \mu_d \end{matrix} \right] \geq 0$$

s.t.

$$B > 0$$

$$\lambda^2 \leq B_1 B_2$$

This relaxation does not change the solution, as the maximization of λ is bound by how large B_1, B_2 can get, so the solution results in $\lambda^2 = B_1 B_2$.

Developing the last constraint:

$$\lambda^2 \leq B_1 B_2 \Leftrightarrow \|\lambda\|^2 \leq B_1 B_2 \stackrel{\text{Fact}}{\Leftrightarrow} \left\| \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \right\|_2 \leq B_1 + B_2 \quad (\Rightarrow)$$

because
 $B_1 > 0, B_2 > 0$

$$\Rightarrow \left\| \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \right\|_2^2 \leq (B_1 + B_2)^2 \Leftrightarrow \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix}^T \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \leq (B_1 + B_2)^2$$

$$\Leftrightarrow \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix}^T \mathbf{1} \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \leq (B_1 + B_2)^2 \stackrel{(\Rightarrow)}{\Leftrightarrow} \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix}^T \frac{1}{(B_1 + B_2)} \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \leq B_1 + B_2 \quad (\Rightarrow)$$

$$\Leftrightarrow \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix}^T \left(\frac{1}{(B_1 + B_2)} \right)^{-1} \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \leq B_1 + B_2 \quad (\Rightarrow) \quad \text{Schur's complement}$$

$$\Leftrightarrow (B_1 + B_2) - \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix}^T \left(\frac{1}{(B_1 + B_2)} \right)^{-1} \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \geq 0 \quad (\Rightarrow)$$

$$\Leftrightarrow \begin{bmatrix} 1(B_1 + B_2) & \begin{bmatrix} 2\lambda \\ B_1 - B_2 \end{bmatrix} \\ \begin{bmatrix} 2\lambda & B_1 - B_2 \end{bmatrix} & (B_1 + B_2) \end{bmatrix} \geq 0$$

→ max λ

B_1, B_2, d, λ

$$\left[\begin{matrix} B_1 d \\ d^T \mu_d \end{matrix} \right] \geq 0$$

s.t.

$$\begin{bmatrix} B > 0 \\ \begin{bmatrix} (B_1 + B_2) & 0 & 2\lambda \\ 0 & (B_1 + B_2)(B_1 - B_2) \end{bmatrix} \\ 2\lambda (B_1 - B_2) (B_1 + B_2) \end{bmatrix} \geq 0$$

Finally, $B > 0$ can be relaxed to $B \geq 0$, as $B_1 = 0 \vee B_2 = 0$ would be an ellipse with an infinite axes length, that cannot be the solution for $S \in \mathbb{R}^2$.

This way, the problem is formulated as an
SDP as:

$$\begin{array}{ll} \text{Max} & \lambda \\ B_1, B_2, d, \lambda & \\ \text{s.t.} & \begin{bmatrix} B & d \\ d^T & \mu_k \end{bmatrix} \succ 0, \quad k = 1, \dots, K \\ & B \succ 0 \end{array}$$

$$\begin{bmatrix} (B_1 + B_2) & 0 & 2\lambda \\ 0 & (B_1 B_2) (B_1 - B_2) \\ 2\lambda & (B_1 - B_2) (B_1 + B_2) \end{bmatrix} \succ 0$$

$$\text{with } B = \begin{bmatrix} B_1 & 0 \\ 0 & B_2 \end{bmatrix}$$

$$\text{and } \mu_k = 1 - P_k^T B P_k + 2 d^T P_k$$

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$$\min_{x_{\text{blend}}} \| Ax_{\text{blend}} - y_{\text{obs}} \|_p$$

$$\text{where } Z = \begin{bmatrix} Z_{\text{blend}} & Z_K \end{bmatrix}$$

$$\text{s.t. } \| x \|_p \leq U$$

$$\text{and } x_{\text{blend}} = \begin{bmatrix} x_{\text{blend}} \\ x_K \end{bmatrix}$$

So we can write

$$\min_{x_{\text{blend}}, x_K} \| Z_{\text{blend}} x_{\text{blend}} + Z_K x_K - y_{\text{obs}} \|_p$$

$$x_{\text{blend}}, x_K$$

$$\text{s.t. } \| x_{\text{blend}} \|_p \leq U$$

$$\| x_K \|_p \leq U$$

To minimize the average error for a given realization of $H_i, i=1, \dots, K$
we do:

$$\min_{x_{\text{blend}}, x_K} \sum_{i=1}^K \| Z_{i, \text{blend}} x_{\text{blend}} + Z_{i, K} x_K - y_{\text{obs}, i} \|_p \cdot P_i, \quad P_i = P(H = H_i)$$

$$\text{s.t. } \| x_{\text{blend}} \|_p \leq U$$

$$\| x_K \|_p \leq U$$

When the objective can be written as:

$$\sum_{i=1}^K \max_{l \in \{1, \dots, m\}} (|Z_{i, l} x_{\text{blend}} + Z_{i, K} x_K - y_{\text{obs}, l}|) \cdot P_i$$

When $Z_{i, l}$, $Z_{i, K}$ and $y_{\text{obs}, l}$ are the l -th line of Z , Z_K and y_{obs} respectively.

Introducing graph variables λ_i , $i = 1, \dots, k$,

$$\lambda_i \geq \max_{l=1, \dots, m^T} (1 \cdot z_{i,l} x_{Blnd} + z_{k,l} x_k - y_{bs,l})$$

The problem becomes:

$$\begin{aligned} & \text{Min}_{x_{Blnd}, x_k, \lambda_1, \dots, \lambda_k} \quad \sum_{i=1}^k \lambda_i p_i \\ \text{s.t.} \quad & \begin{aligned} & u x_{Blnd} u \in U \\ & u x_k u \in U \\ & \lambda_i \geq \max_{l=1, \dots, m^T} (1 \cdot z_{i,l} x_{Blnd} + z_{k,l} x_k - y_{bs,l}), \text{ for } i = 1, \dots, k \end{aligned} \end{aligned} \quad (\rightarrow)$$

$$\begin{aligned} (\rightarrow) \quad & \text{Min}_{x_{Blnd}, x_k, \lambda_1, \dots, \lambda_k} \quad \sum_{i=1}^k \lambda_i p_i \\ & \cup \geq \max_{j=1, \dots, m^S} (1 \cdot x_{Blnd,j}) \quad (\rightarrow) \\ & \cup \geq \max_{j=1, \dots, m^{(T-S)}} (1 \cdot x_k j) \\ & \lambda_i \geq \max_{l=1, \dots, m^T} (1 \cdot z_{i,l} x_{Blnd} + z_{k,l} x_k - y_{bs,l}), \text{ for } i = 1, \dots, k \end{aligned}$$

$$\begin{aligned} (\rightarrow) \quad & \text{Min}_{x_{Blnd}, x_k, \lambda_1, \dots, \lambda_k} \quad \sum_{i=1}^k \lambda_i p_i \\ & \cup \geq x_{Blnd,j}, \text{ for } j = 1, \dots, m^S \\ & \cup \geq -x_{Blnd,j}, \text{ for } j = 1, \dots, m^S \\ & \cup \geq x_k j, \text{ for } j = 1, \dots, m^{(T-S)} \\ & \cup \geq -x_k j, \text{ for } j = 1, \dots, m^{(T-S)} \\ & \lambda_i \geq (z_{i,l} x_{Blnd} + z_{k,l} x_k - y_{bs,l}), \text{ for } i = 1, \dots, k; l = 1, \dots, m^T \\ & \lambda_i \geq -(z_{i,l} x_{Blnd} + z_{k,l} x_k - y_{bs,l}), \text{ for } i = 1, \dots, k; l = 1, \dots, m^T \end{aligned}$$

Min

$x_{\text{Blend}}, \lambda_K, \lambda_1, \dots, \lambda_N$

$$\sum_{i=1}^K \lambda_i p_i \rightarrow \text{linear function of } (\lambda_1, \dots, \lambda_N)$$

$$U \geq x_{\text{Blend}} + \text{for } j=1, \dots, m^S$$

$$U \geq -x_{\text{Blend}} \text{ for } j=1, \dots, m^S$$

$$U \geq x_{Kj} \text{ for } j=1, \dots, m(T-S)$$

$$U \geq -x_{Kj} \text{ for } j=1, \dots, m(T-S)$$

$$\lambda_i \geq (2z_{i1} x_{\text{Blend}} + 2z_{i2} x_N - y_{B,i}) \text{ for } i=1, \dots, K; l=1, \dots, m^T$$

$$\lambda_i \geq -(2z_{iL} x_{\text{Blend}} + 2z_{i2} x_N - y_{B,i}) \text{ for } i=1, \dots, K; l=1, \dots, m^T$$

$Z(m+Km)^T$
Affine inequalities on the
optimization variables

This is the formulation
of the problem as an LP.