## **Examples of LMIs**

SIMONE NALDI, Université de Limoges, CNRS, XLIM, France

MOHAB SAFEY EL DIN, Sorbonne Université, CNRS, LIP6, France

ADRIEN TAYLOR, Inria, École normale supérieure, PSL Research University, France

WEIJIA WANG, Sorbonne Université, CNRS, LIP6, France

In this document, we provide explicit examples of LMIs that are considered in our work. For their context and motivation, we refer to the documents provided in the same repository.

• MKN11 denotes the LMI defined by

$$A = \begin{pmatrix} \varepsilon & \frac{1}{2} + \frac{3\varepsilon}{2} & 0 & 0 & -x_{24} & 0\\ \frac{1}{2} + \frac{3\varepsilon}{2} & 1 + 3\varepsilon & 0 & x_{24} & 0 & 0\\ 0 & 0 & \varepsilon & 0 & 0 & 0\\ 0 & x_{24} & 0 & 3\varepsilon & -\frac{3}{2} + 3\varepsilon & \frac{3\varepsilon}{2}\\ -x_{24} & 0 & 0 & -\frac{3}{2} + 3\varepsilon & 3\varepsilon & \frac{3\varepsilon}{2}\\ 0 & 0 & 0 & \frac{3\varepsilon}{2} & \frac{3\varepsilon}{2} & 1 + \varepsilon \end{pmatrix}$$

$$\mathbf{y} = \varepsilon, \quad \mathbf{x} = \mathbf{x}_{24}.$$

• RBN11 denotes the LMI defined by

$$A = \begin{pmatrix} 1 & -\frac{1}{2} + 2\epsilon & 0 & 0 & -x_{24} & 0 \\ -\frac{1}{2} + 2\epsilon & -1 + 4\epsilon & 0 & x_{24} & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & x_{24} & 0 & -1 + 4\epsilon & \frac{3}{2} + \frac{3}{2}\epsilon & -\frac{1}{2} + 2\epsilon \\ -x_{24} & 0 & 0 & \frac{3}{2} + \frac{3}{2}\epsilon & -1 + 4\epsilon & -\frac{1}{2} + 2\epsilon \\ 0 & 0 & 0 & -\frac{1}{2} + 2\epsilon & -\frac{1}{2} + 2\epsilon & 1 \end{pmatrix}$$

$$\mathbf{y} = \varepsilon, \quad \mathbf{x} = \mathbf{x}_{24}.$$

• GRD24 denotes the LMI defined by

$$A = \begin{pmatrix} \frac{\mu L(\lambda_1 + \lambda_3 + \lambda_5 + \lambda_6)}{L - \mu} & \star & \star \\ -\frac{\lambda_5 \mu + L(\gamma \mu (\lambda_1 + \lambda_6) + \lambda_3)}{L - \mu} & \frac{\lambda_2 + \lambda_3 + \lambda_4 + \lambda_5 + \gamma \mu (\gamma L(\lambda_1 + \lambda_2 + \lambda_4 + \lambda_6) - 2\lambda_2) - 2\gamma \lambda_4 L}{L - \mu} & \star \\ -\frac{\lambda_6 \mu + \lambda_1 L}{L - \mu} & \frac{\gamma \lambda_4 \mu + \gamma \lambda_6 \mu - \lambda_2 - \lambda_4 + \gamma L(\lambda_1 + \lambda_2)}{L - \mu} & \frac{\lambda_1 + \lambda_2 + \lambda_4 + \lambda_6}{L - \mu} \end{pmatrix},$$

and  $\boldsymbol{y}=(\mu,\tau), \, \boldsymbol{x}=(\lambda_1,\dots,\lambda_6),$  with the constraints  $\lambda_1,\dots,\lambda_6\geq 0, \, \gamma=\frac{2}{L+\mu}, \, L=1, \, 0<\mu<1,$  and  $\begin{cases} -\lambda_2+\lambda_3+\lambda_4-\lambda_5+\tau &=0\\ \lambda_1+\lambda_2-\lambda_4-\lambda_6-1 &=0, \end{cases}$ 

$$\begin{cases} -\lambda_2 + \lambda_3 + \lambda_4 - \lambda_5 + \tau &= 0\\ \lambda_1 + \lambda_2 - \lambda_4 - \lambda_6 - 1 &= 0, \end{cases}$$

so that the LMI has 4 free variables, say  $\lambda_1, \lambda_4, \lambda_5, \lambda_6$ , and 2 parameters

Authors' Contact Information: Simone Naldi, Université de Limoges, CNRS, XLIM, Limoges, France; Mohab Safey El Din, Sorbonne Université, CNRS, LIP6, Paris, France; Adrien Taylor, Inria, École normale supérieure, PSL Research University, Paris, France; Weijia Wang, Sorbonne Université, CNRS, LIP6, Paris, France,

- GRD23 denotes the LMI obtained by setting  $\lambda_6 = 0$  in GRD24.
- GRD14 denotes the LMI obtained by moving  $\tau$  from  $\boldsymbol{y}$  to  $\boldsymbol{x}$  in GRD21.
- GRD22 denotes the LMI obtained by setting  $\lambda_5 = 0$  in GRD23.
- GRD13 denotes the LMI obtained by moving  $\tau$  from y to x in GRD21.
- GRD21 denotes the LMI obtained by setting  $\lambda_4 = 0$  in GRD22.
- GRD12 denotes the LMI obtained by moving  $\tau$  from  $\boldsymbol{y}$  to  $\boldsymbol{x}$  in GRD21.
- PPM31 denotes the LMI defined by

$$A = \begin{pmatrix} 2\lambda\mu + 2\tau - 2 & -\gamma(2\lambda\mu - 2) - \lambda \\ -\gamma(2\lambda\mu - 2) - \lambda & \gamma(\gamma(2\lambda\mu - 2) + 2\lambda) \end{pmatrix},$$

and  $y = (\mu, \gamma, \tau)$ ,  $x = \lambda$ , with the constraints  $\mu > 0$ ,  $\gamma > 0$ ,  $\tau > 0$ .

- PPM21 denotes the LMI obtained by setting  $\gamma = 1$  in PPM31.
- DRS42 denotes the LMI proposed in [1, SM3.1.1.], defined by

$$A = \begin{pmatrix} \rho^2 + \beta \lambda_\beta^B - 1 & -\theta + \frac{\lambda_\mu^A}{2} & \theta - (\frac{1}{2} + \beta) \lambda_\beta^B \\ -\theta + \frac{\lambda_\mu^A}{2} & -\theta^2 + (1 + \mu) \lambda_\mu^A & \theta^2 - \lambda_\mu^A \\ \theta - (\frac{1}{2} + \beta) \lambda_\beta^B & \theta^2 - \lambda_\mu^A & -\theta^2 + (1 + \beta) \lambda_\beta^B \end{pmatrix},$$

and  $\mathbf{y} = (\mu, \beta, \rho, \theta)$ ,  $\mathbf{x} = (\lambda_{\mu}^{A}, \lambda_{\beta}^{B})$ , with the constraints  $\mu > 0$ ,  $0 < \theta < 2$ ,  $\beta - \mu \ge 0$ .

- DRS32 denotes the LMI obtained by setting  $\theta$  = 1 in DRS42.
- DRS43 denotes the LMI proposed in [1, SM3.2.2.], defined by

$$A = \begin{pmatrix} \rho^2 + \lambda_L^B - 1 & \frac{\lambda_\mu^A}{2} - \theta & \theta - \lambda_L^B - \frac{\lambda_\mu^B}{2} \\ \frac{\lambda_\mu^A}{2} - \theta & -\theta^2 + \lambda_\mu^A + \lambda_\mu^A \mu & \theta^2 - \lambda_\mu^A \\ \theta - \lambda_L^B - \frac{\lambda_\mu^B}{2} & \theta^2 - \lambda_\mu^A & -\lambda_L^B L^2 - \theta^2 + \lambda_L^B + \lambda_\mu^B \end{pmatrix},$$

and  $\mathbf{y} = (\mu, L, \rho, \theta), \mathbf{x} = (\lambda_{\mu}^A, \lambda_{\mu}^B, \lambda_{\mu}^B), \text{ with } \mu > 0, L - \mu > 0.$ 

• DRS33 denotes the LMI obtained by setting  $\theta = 1$  in DRS43.

## References

[1] E. K. Ryu, A. B. Taylor, C. Bergeling, and P. Giselsson. Operator splitting performance estimation: tight contraction factors and optimal parameter selection. SIAM Journal on Optimization, 30(3):2251–2271, 2020.