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1. **Schur complement lemma:** Find an equivalent semidefinite condition of the form $G(x) \succeq 0$ for each of the following statements. Make sure the matrix $G(x)$ you obtained is affine w.r.t. x , where x is a vector or matrix variable of appropriate dimension. Please show your steps.
 - (a) (Singular value bound): $\sigma(A(x)) < \beta$, where $A : \mathbb{R}^n \rightarrow \mathbb{R}^{q \times m}$ is affine in $x \in \mathbb{R}^n$ and $\sigma(\cdot)$ denotes the singular value of a matrix.
 - (b) (Riccati inequality): $A^T x + xA + xBR^{-1}B^T x + Q \prec 0$, with $x \in \mathcal{S}_{++}^n$, $R \in \mathcal{S}_{++}^p$, $Q \in \mathcal{S}^n$ and $A \in \mathbb{R}^{n \times n}$, and $B \in \mathbb{R}^{n \times p}$.
2. **Ellipsoid:** Ellipsoid in \mathbb{R}^n have two equivalent representations: (i) $E_1(P, x_c) = \{x \in \mathbb{R}^n : (x - x_c)^T P^{-1} (x - x_c) \leq 1\}$ and (ii) $E_2(A, x_c) = \{Au + x_c : \|u\|^2 \leq 1\}$. The second representation can be derived from the first by letting $A = P^{1/2}$. Given $E_1(P, x_c)$ with $P \in \mathcal{S}_{++}^n$, its volume is $\nu_n \sqrt{\det(P)}$ where ν_n is the volume of unit ball in \mathbb{R}^n , its semi-axes directions are given by the eigenvectors of P and the lengths of semi-axes are $\sqrt{\lambda_i}$, where λ_i are eigenvalues of P .
 - (a) Given a half space $\{x \in \mathbb{R}^n : a^T x \leq 1\}$. Show that the Ellipsoid $E_2(A, 0)$ is contained in the half space if and only if $a^T A A^T a \leq 1$.
 - (b) Note that for any $P \in \mathcal{S}_{++}^n$, the function $\log(\det(P))$ is concave in the matrix variable P . Formulate a convex optimization problem to find the matrix $P \in \mathcal{S}_{++}^n$ such that $E_1(P, 0)$ is the largest ellipsoid contained in the polyhedron $\{x \in \mathbb{R}^n : a_i^T x \leq 1, i = 1, \dots, m\}$
 - (c) Use Drake to solve the above problem with $a_1^T = [-1, 1]$, $a_2^T = [2, -1]$, $a_3^T = [1, 3]$, $a_4^T = [-2, -5]$. Visualize the polyhedron region and your ellipsoid solution (you can use Matlab for the visualization if you prefer matlab)

3. **Stability of Lur'e system:** Consider a nonlinear system

$$\dot{x} = Ax + b\phi(t, c^T x)$$

where $A \in \mathbb{R}^{n \times n}$, $b \in \mathbb{R}^n$, $c \in \mathbb{R}^n$, and for each time t , the function $\phi(t, \cdot) : \mathbb{R} \rightarrow \mathbb{R}$ satisfies sector nonlinearity $|\phi(t, y)| \leq \alpha|y|$ for all y (but is otherwise unknown). Such a system represents a very general class of control systems involving time-varying nonlinearities and/or uncertainties, and is often called a *Lur'e* problem.

We would like to find a positive definite Lyapunov function $V(x) = x^T P x$ that satisfies $\dot{V}(x) \leq -\beta V(x)$ for all x , and for any function ϕ satisfying the inequality given above. You can assume that A, b, c, α, β are given.

- (a) Explain how to find such a P (or determine that no such P exists) by expressing the problem as an LMI.
- (b) Use CVX to construct such a Lyapunov function for the following instance of Lur'e problem:

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & -3 & -3 \end{bmatrix}, b = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, c = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \alpha = 0.7, \beta = 0.1$$

4. **Stabilization via LMIs:** Consider the time-varying LDS (linear dynamical system)

$$\dot{x}(t) = A(t)x(t) + Bu(t),$$

with $x(t) \in \mathbb{R}^n$ and $u(t) \in \mathbb{R}^m$, where $A(t) \in \{A_1, \dots, A_M\}$. Thus, the dynamics matrix $A(t)$ can take any of M values, at any time. We seek a linear state feedback gain matrix $K \in \mathbb{R}^{m \times n}$ for which the closed-loop system

$$\dot{x}(t) = [A(t) + BK]x(t),$$

is globally asymptotically stable. But even if you are given a specific state feedback gain matrix K , this is very hard to determine. So we will require the existence of a quadratic Lyapunov function that establishes exponential stability of the closed-loop system, *i.e.*, a matrix $P = P^T \succ 0$ for which

$$\dot{V}(z, t) = z^T [(A(t) + BK)^T P + P(A(t) + BK)] z \leq -\beta V(z)$$

for all z , and for any possible value of $A(t)$. (The parameter $\beta > 0$ is given, and sets a minimum decay rate for the closed-loop trajectories.)

So roughly speaking we seek

- a stabilizing state feedback gain, and
- a quadratic Lyapunov function that certifies the closed-loop performance.

In this problem, you will use LMIs to find both K and P , *simultaneously*.

- Pose the problem of finding P and K as an LMI problem. *Hint:* Starting from the inequality above, you will not get an LMI in the variables P and K (although you will have a set of matrix inequalities that are affine in K , for fixed P , and linear in P , for fixed K). Use the new variables $X = P^{-1}$ and $Y = KP^{-1}$. Be sure to explain why you can change variables.
- Carry out your method for the specific problem instance

$$A_1 = \begin{bmatrix} -0.5 & 0.3 & 0.4 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, \quad A_2 = \begin{bmatrix} -0.7 & 0.1 & -0.2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, \quad A_3 = \begin{bmatrix} 0.6 & -0.7 & 0.2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix},$$

$B = (1, 0, 0)$, and $\beta = 1$. (Thus, we require a closed-loop decay at least as fast as $e^{-t/2}$.)

5. **Derivation of Dual SDP (Optional; will not be graded):** Consider the prime and dual SDP problems discussed in class.

$$(\text{PSDP}) : \begin{cases} \min_X & C \bullet X \\ \text{subj.} & A_i \bullet X = b_i, \\ & i = 1, \dots, m \\ & X \succeq 0 \end{cases} \quad (\text{DSDP}) : \begin{cases} \max_{y \in \mathbb{R}^m} & b^T y \\ \text{subj.} & \sum_{i=1}^m y_i A_i \preceq C \end{cases}$$

This homework problem intends to guide you through the derivation of the dual SDP problem. We need to introduce a few definitions. In general, the Lagrangian dual of a constrained optimization problem $\min_{x \in \mathcal{X}} \{f(x) : h(x) = 0\}$ is defined as

$$\max_{y \in \mathbb{R}^m} \left(\min_{x \in \mathcal{X}} \{f(x) + y^T h(x)\} \right)$$

To derive the desired dual form, we also need a concept: *dual cone*. Let \mathcal{V} be an inner product space. Suppose $K \subseteq \mathcal{V}$ is a cone, i.e., $\lambda x \in K$ for all $x \in K$. The dual cone of K is defined as

$$Dual(K) = \{z \in \mathcal{V} : \langle z, x \rangle \geq 0, \forall x \in K\}$$

The dual of the PSD cone is thus $dual(\mathcal{S}_+^n) = \{Y \in \mathcal{S}^n : Y \bullet X \geq 0, \forall X \in \mathcal{S}_+^n\}$. A well-known fact about PSD cone is that it is self-dual, i.e. $\mathcal{S}_+^n = dual(\mathcal{S}_+^n)$.

- (a) Write down the Lagrangian dual problem for (PSDP).
- (b) Show that $\min_{X \succeq 0} \langle Y, X \rangle = \begin{cases} 0 & \text{if } Y \succeq 0 \\ -\infty & \text{otherwise} \end{cases}$ (Hint: use the fact that \mathcal{S}_+^n is acute and self-dual).
- (c) Show that the Lagrangian dual problem of (PSDP) can be reduced to the form (DSDP) given above.