- 1. Is the matrix $A^T A$ positive semi-definite? (hint, for $A^T A$ to be positive semi-definite, $x^T A^T A x \ge 0$ for all x.
 - (a) yes
 - (b) no
- 2. Is the following linear system controllable?

$$x_{k+1} = \begin{bmatrix} 1 & 3\\ -2 & 0 \end{bmatrix} x_k + \begin{bmatrix} 0\\ .1 \end{bmatrix} u_k \tag{1}$$

(a) yes

- (b) no
- 3. Below is a finite-horizon LQR problem,

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$$\min_{x_{1:N},u_{1:N-1}} \sum_{i=1}^{N-1} \left[\frac{1}{2} x_i^T Q x_i + \frac{1}{2} u_i^T R u_i \right] + \frac{1}{2} x_N^T Q_f x_N$$
(2)

$$x_1 = x_{\rm IC},\tag{3}$$

$$x_{i+1} = Ax_i + Bu_i \quad \text{for } i = 1, 2, \dots, N-1,$$
(4)

where $Q \succeq 0$, $R \succ 0$, and [A, B] is controllable. We can solve this problem with either convex optimization that solves for (x, u) directly, or a Ricatti recursion that solves for an optimal policy $u_i = -K_i x_i$. Will these two solutions be equivalent?

- (a) yes
- (b) no
- 4. If we modify the finite-horizon LQR problem to incorporate bound constraints on the controls, we get the following problem:

$$\min_{x_{1:N},u_{1:N-1}} \sum_{i=1}^{N-1} \left[\frac{1}{2} x_i^T Q x_i + \frac{1}{2} u_i^T R u_i \right] + \frac{1}{2} x_N^T Q_f x_N$$
(5)

st
$$x_1 = x_{\rm IC},$$
 (6)

$$x_{i+1} = Ax_i + Bu_i \quad \text{for } i = 1, 2, \dots, N-1,$$
(7)

 $u_{min} \le u_i \le u_{max}$ for $i = 1, 2, \dots, N - 1.$ (8)

If we take the Ricatti solution $(K_{1:N-1})$ from the original finite-horizon LQR problem, and apply the following policy:

$$u_i = \max(u_{min}, \min(-K_i x_i, u_{max})).$$
(9)

Will we recover a solution to the new optimization problem? You should think about the theory as well as implementing it yourself and trying. First figure out what $\max(a, \min(u, b))$ does when a < b, making sure to use max.() and min.() if you're doing this in Julia.

- (a) yes
- (b) no

- 5. Which of the following cost functions are convex in $x \in \mathbb{R}^N$? $(Q \succ 0)$
 - (a) $c^T x$
 - (b) $-c^T x$
 - (c) $\sum_i x_i$
 - (d) $||x||_1$
 - (e) $||x||_2$
 - (f) $||x||_2^2$
 - (g) $||Ax b||_1$
 - (h) $x^T Q x$
 - (i) $(Ax b)^T Q(Ax b)$

6. Which of the following constraints are convex in $x \in \mathbb{R}^N$? $(Q \succ 0)$

- (a) Ax = b
- (b) $Ax \leq b$
- (c) $||x||_2 \ge 3$
- (d) $||Ax b||_1 \le 3$
- (e) $||Ax b||_2 \le 3$
- (f) $||Ax b||_2 = 3$
- (g) $||x||_2^2 \le 3$
- (h) $x^T Q x \leq 3$
- (i) $x^T Q x = 3$
- (j) $x^T Q x \ge 3$
- (k) $x^T Q x + q^T x \le 3$