

LQR

AE 353

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Bretl

$$\dot{x} = Ax + Bu$$

$$u = -Kx$$

↑
HOW TO FIND K?

① Gain tuning (i.e., guess and check)

- make a small change to K
- check if all eigenvalues of $A - BK$ have negative real part
- repeat until satisfied

② Eigenvalue placement

- choose desired eigenvalue locations
- apply "place-poles" or Ackermann's method

③ LQR (minimize a cost)

- choose weights on the cost of non-zero x and u
- choose K to minimize total, integrated cost

EXAMPLE (RUN DEMO)

$$\begin{aligned} \dot{x} &= \overset{A}{\downarrow} [5] x + \overset{B}{\downarrow} [1] u \\ u &= - \overset{K}{\uparrow} [k] x \end{aligned} \quad \left. \vphantom{\begin{aligned} \dot{x} &= [5] x + [1] u \\ u &= -[k] x \end{aligned}} \right\} \dot{x} = \overset{A-BK}{\downarrow} [5-k] x$$

$k > 5$ for stability

$k = 5 - p$ for eigenvalue at p

Linear Quadratic Regulator (LQR)

minimize $u [t_0, \infty)$
subject to

$$\int_{t_0}^{\infty} (x(t)^T Q x(t) + u(t)^T R u(t)) dt$$

Cost

$$\dot{x}(t) = A x(t) + B u(t)$$
$$x(t_0) = x_0$$

total cost

constraints

The minimizer (i.e., the input that achieves minimum cost) is

$$u(t) = -K x(t)$$

and the minimum (i.e., the minimum cost) is

$$x_0^T P x_0$$

where K and P can be found in python as follows:

```
def lqr(A, B, Q, R):  
    P = linalg.solve_continuous_are(A, B, Q, R)  
    K = linalg.inv(R) @ B.T @ P  
    return K, P
```

$$\begin{array}{l} \text{minimize} \\ u [t_0, \infty), x [t_0, \infty) \\ \text{subject to} \end{array} \quad \int_{t_0}^{\infty} (x(t)^T Q x(t) + u(t)^T R u(t)) dt$$
$$\begin{array}{l} \dot{x}(t) = A x(t) + B u(t) \\ x(t_0) = x_0 \end{array}$$

Why is the cost "quadratic" and what does it really mean?

$$x = [x_1] \quad Q = [q_1] \quad u = [u_1] \quad R = [r_1]$$

$$x^T Q x + u^T R u = [x_1] [q_1] [x_1] + [u_1] [r_1] [u_1]$$

$$= [q_1 x_1^2] + [r_1 u_1^2]$$

$$= [q_1 x_1^2 + r_1 u_1^2]$$

Why is the cost "quadratic" and what does it really mean?

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$u = [u_1]$$

$$Q = \begin{bmatrix} q_1 & q_3 \\ q_3 & q_2 \end{bmatrix}$$

$$R = [r_1]$$

$$x^T Q x + u^T R u = [x_1 \quad x_2] \begin{bmatrix} q_1 & q_3 \\ q_3 & q_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + [u_1] [r_1] [u_1]$$

$$= [x_1 \quad x_2] \begin{bmatrix} q_1 x_1 + q_3 x_2 \\ q_3 x_1 + q_2 x_2 \end{bmatrix} + [r_1 u_1^2]$$

$$= [q_1 x_1^2 + q_3 x_1 x_2 + q_3 x_1 x_2 + q_2 x_2^2] + [r_1 u_1^2]$$

$$= [q_1 x_1^2 + 2q_3 x_1 x_2 + q_2 x_2^2 + r_1 u_1^2]$$

What Q and R would produce a given cost?

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$u = [u_1]$$

$$Q = \begin{bmatrix} q_1 & q_3 \\ q_3 & q_2 \end{bmatrix}$$

$$R = [r_1]$$

$$[q_1 x_1^2 + 2q_3 x_1 x_2 + q_2 x_2^2 + r_1 u_1^2]$$

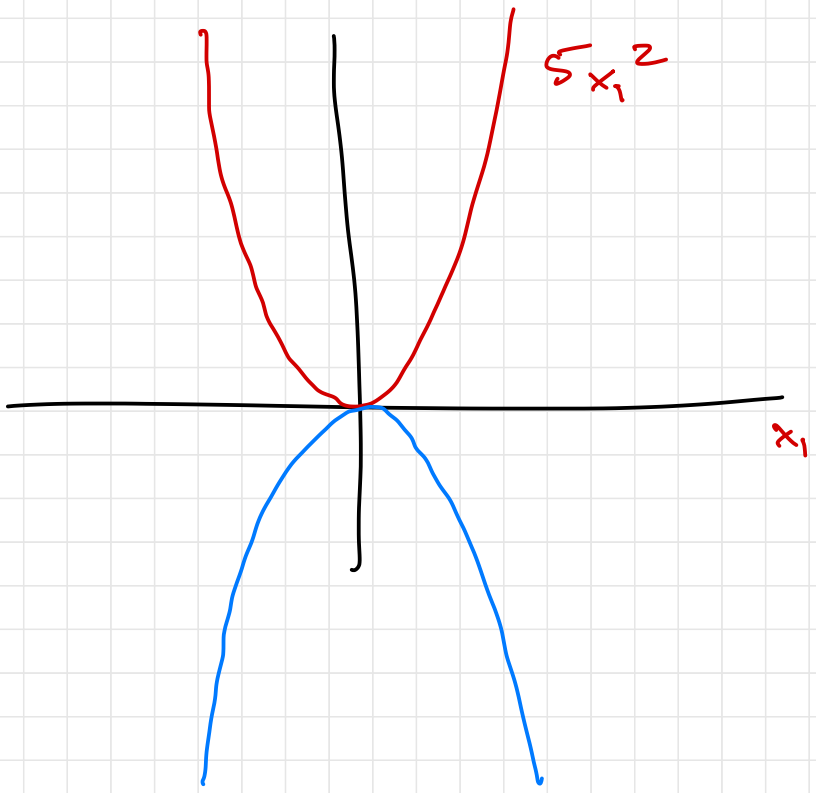
$$x^T Q x + u^T R u = 5x_1^2 - 2x_1 x_2 + 2x_2^2 + 1u_1^2$$

↓

$$Q = \begin{bmatrix} 5 & -1 \\ -1 & 2 \end{bmatrix}$$

$$R = [1]$$

$$\sqrt{x_1^2}$$



Q and R are commonly chosen to be diagonal

$$Q = \text{diag}(q_1, \dots, q_{n_x})$$

↙ all positive

$$R = \text{diag}(r_1, \dots, r_{n_u})$$

↑
all positive

Rule of thumb ("Bryson's rule")

$$q_i = \left(\frac{1}{X_{i, \max}} \right)^2$$

$$r_i = \left(\frac{1}{u_{i, \max}} \right)^2$$