

Linear-Quadratic Control

Jenny Hong Nicholas Moehle Stephen Boyd

ENGR108
Stanford University

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Outline

Linear dynamical system

Control

Variations

Examples

Linear quadratic regulator

Linear dynamical system

$$x_{t+1} = Ax_t + Bu_t, \quad t = 1, 2, \dots$$

- ▶ n -vector x_t is *state* at time t
- ▶ m -vector u_t is *input* at time t
- ▶ $n \times n$ matrix A is *dynamics matrix*
- ▶ $n \times m$ matrix B is *input matrix*
- ▶ sequence x_1, x_2, \dots is called *state trajectory*

Simulation

- ▶ given x_1, u_1, u_2, \dots find x_2, x_3, \dots
- ▶ can be done by recursion: for $t = 1, 2, \dots$,

$$x_{t+1} = Ax_t + Bu_t$$

Vehicle example

consider a vehicle moving in a plane:

- ▶ sample position and velocity at times $\tau = 0, h, 2h, \dots$
- ▶ 2-vectors p_t and v_t are position and velocity at time ht
- ▶ 2-vector u_t gives applied force on the vehicle time ht
- ▶ friction force is $-\eta v_t$
- ▶ vehicle has mass m
- ▶ for small h ,

$$m \frac{v_{t+1} - v_t}{h} \approx -\eta v_t + u_t, \quad \frac{p_{t+1} - p_t}{h} \approx v_t$$

- ▶ we use approximate state update

$$v_{t+1} = (1 - h\eta/m)v_t + (h/m)u_t, \quad p_{t+1} = p_t + hv_t$$

- ▶ vehicle state is 4-vector $x_t = (p_t, v_t)$
- ▶ dynamics recursion is

$$x_{t+1} = Ax_t + Bu_t,$$

where

$$A = \begin{bmatrix} 1 & 0 & h & 0 \\ 0 & 1 & 0 & h \\ 0 & 0 & 1 - h\eta/m & 0 \\ 0 & 0 & 0 & 1 - h\eta/m \end{bmatrix}, \quad B = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ h/m & 0 \\ 0 & h/m \end{bmatrix}$$

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- ▶ x_1 is given
- ▶ choose u_1, u_2, \dots, u_{T-1} to achieve some goals, e.g.,
 - terminal state should have some fixed value: $x_T = x^{\text{des}}$
 - u_1, u_2, \dots, u_{T-1} should be small, say measured as

$$\|u_1\|^2 + \dots + \|u_{T-1}\|^2$$

(sometimes called 'energy')

- ▶ many control problems are linearly constrained least-squares problems

Minimum-energy state transfer

- ▶ given initial state x_1 and desired final state x^{des}
- ▶ choose u_1, \dots, u_{T-1} to minimize 'energy'

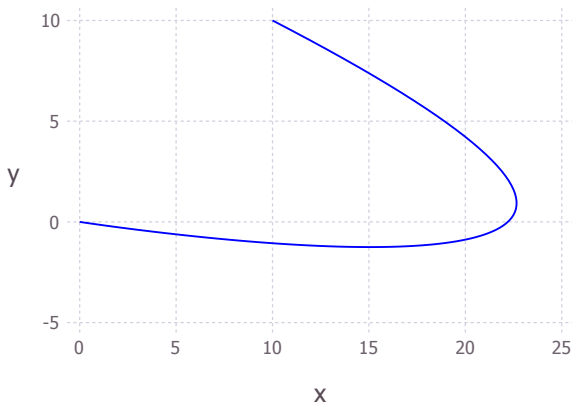
$$\begin{array}{ll}\text{minimize} & \|u_1\|^2 + \dots + \|u_{T-1}\|^2 \\ \text{subject to} & x_{t+1} = Ax_t + Bu_t, \quad t = 1, \dots, T-1 \\ & x_T = x^{\text{des}}\end{array}$$

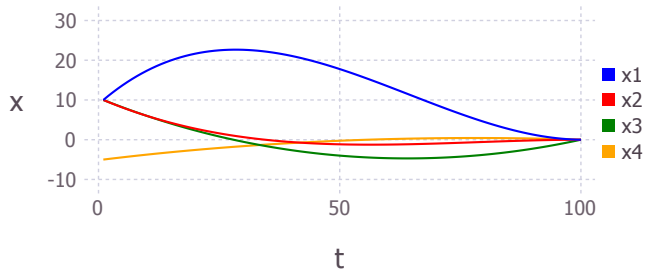
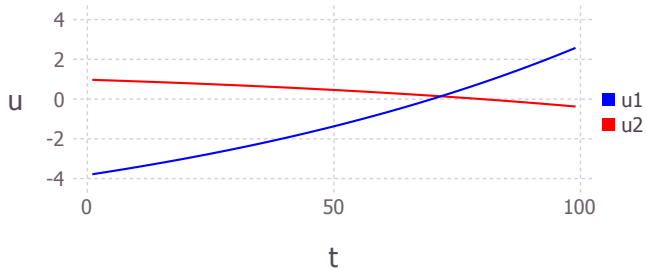
variables are $x_2, \dots, x_T, u_1, \dots, u_{T-1}$

- ▶ roughly speaking: find minimum energy inputs that steer the state to given target state over T periods

State transfer example

vehicle model with $T = 100$, $x_1 = (10, 10, 10, -5)$, $x^{\text{des}} = 0$





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Output tracking

- ▶ $y_t = Cx_t$ is output (e.g., position)
- ▶ y_t should follow a desired trajectory, i.e., sum square *tracking error*

$$\|y_2 - y_2^{\text{des}}\|^2 + \cdots + \|y_T - y_T^{\text{des}}\|^2$$

should be small

- ▶ the output tracking problem is

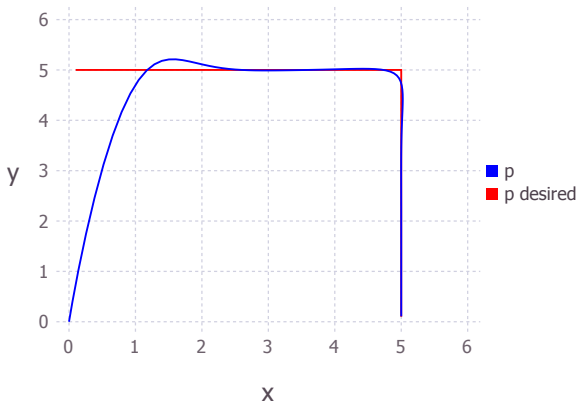
$$\begin{array}{ll} \text{minimize} & \sum_{t=2}^T \|y_t - y_t^{\text{des}}\|^2 + \rho \sum_{t=1}^{T-1} \|u_t\|^2 \\ \text{subject to} & x_{t+1} = Ax_t + Bu_t, \quad t = 1, \dots, T-1 \\ & y_t = Cx_t, \quad t = 1, \dots, T-1 \end{array}$$

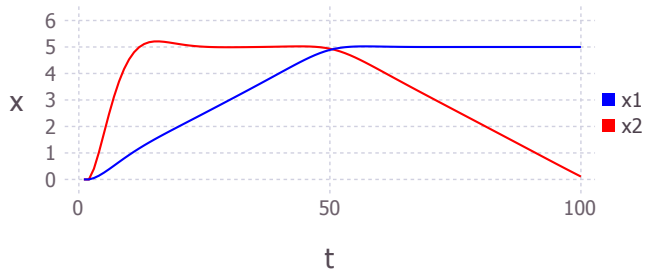
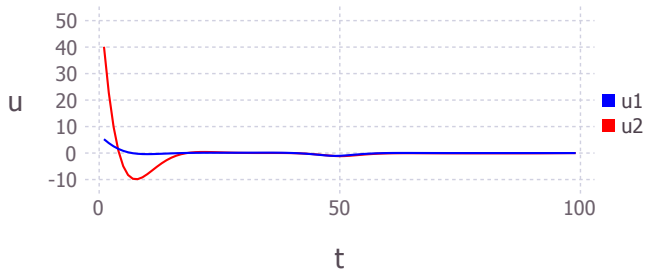
variables are $x_2, \dots, x_T, u_1, \dots, u_{T-1}, y_2, \dots, y_T$

- ▶ parameter $\rho > 0$ trades off control ‘energy’ and tracking error

Output tracking example

vehicle model with $T = 100$, $\rho = 0.1$, $x_1 = 0$, $y_t = p_t$ (position tracking)





Waypoints

- ▶ using output, can specify *waypoints*
- ▶ specify output (position) $w^{(k)}$ at time t_k at K total places

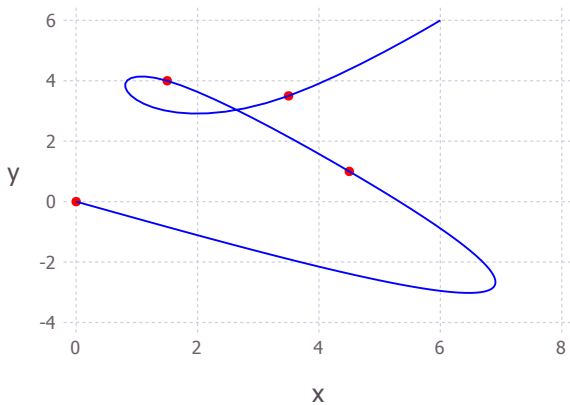
$$\begin{array}{ll}\text{minimize} & \|u_1\|^2 + \cdots + \|u_{T-1}\|^2 \\ \text{subject to} & x_{t+1} = Ax_t + Bu_t, \quad t = 1, \dots, T-1 \\ & Cx_{t_k} = w^{(k)}, \quad k = 1, \dots, K\end{array}$$

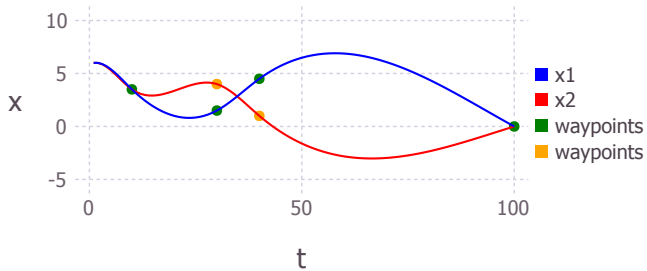
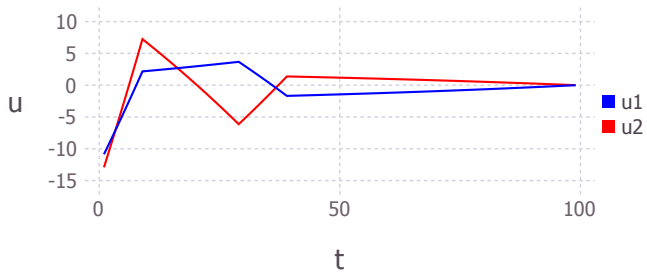
variables are $x_2, \dots, x_T, u_1, \dots, u_{T-1}$

Waypoints example

- ▶ vehicle model
- ▶ $T = 100$, $x_1 = (10, 10, 20, 0)$, $x^{\text{des}} = 0$
- ▶ $K = 4$, $t_1 = 10$, $t_2 = 30$, $t_3 = 40$, $t_4 = 80$
- ▶ $w^{(1)} = \begin{bmatrix} 3.5 \\ 3.5 \end{bmatrix}$, $w^{(2)} = \begin{bmatrix} 1.5 \\ 4 \end{bmatrix}$, $w^{(3)} = \begin{bmatrix} 4.5 \\ 1 \end{bmatrix}$, $w^{(4)} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Waypoints example





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Rendezvous

- ▶ we control two vehicles with dynamics

$$x_{t+1} = Ax_t + Bu_t, \quad z_{t+1} = Az_t + Bv_t$$

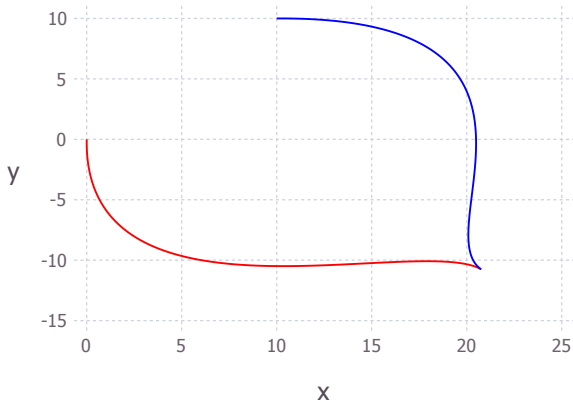
- ▶ final relative state constraint $x_T = z_T$
- ▶ formulate as state transfer problem:

$$\begin{aligned} & \text{minimize} && \sum_{t=1}^{T-1} (\|u_t\|^2 + \|v_t\|^2) \\ & \text{subject to} && x_{t+1} = Ax_t + Bu_t, \quad t = 1, \dots, T-1, \\ & && z_{t+1} = Az_t + Bv_t, \quad t = 1, \dots, T-1, \\ & && x_T = z_T \end{aligned}$$

variables are $x_2, \dots, x_T, u_1, \dots, u_{T-1}, z_2, \dots, z_T, v_1, \dots, v_{T-1}$

Rendezvous example

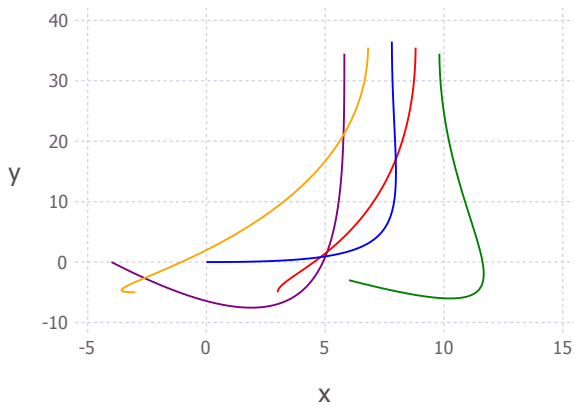
$$x_1 = (0, 0, 0, -5), z_1 = (10, 10, 5, 0)$$



Formation

- ▶ generalize rendezvous example to several vehicles
- ▶ final position for each vehicle defined relative to others (e.g., relative to a 'leader')
- ▶ leader has a final velocity constraint

Formation example



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Linear quadratic regulator

- ▶ minimize energy while driving state to the origin:

$$\begin{array}{ll}\text{minimize} & \sum_{t=2}^T \|x_t\|^2 + \rho \sum_{t=1}^{T-1} \|u_t\|^2 \\ \text{subject to} & x_{t+1} = Ax_t + Bu_t, \quad t = 1, \dots, T-1\end{array}$$

variables are $x_2, \dots, x_T, u_1, \dots, u_{T-1}$

- ▶ $\sum_{t=2}^T \|x_t\|^2$ is (sum square) *regulation*
- ▶ $x = 0$ is some desired (equilibrium, target) state
- ▶ parameter $\rho > 0$ trades off regulation versus control ‘energy’

- ▶ LQR problem is a linearly constrained least-squares problem:

$$\begin{array}{ll}\text{minimize} & \|Fz\|^2 \\ \text{subject to} & Gz = d\end{array}$$

- ▶ variable z is $(x_2, \dots, x_T, u_1, \dots, u_{T-1})$
- ▶ F, G depend on A, B, ρ ; d depends (linearly) on x_1
- ▶ solution is $\hat{z} = Hd$ for some H
- ▶ optimal first input \hat{u}_1 is a linear function of x_1 , i.e.,

$$\hat{u}_1 = Kx_1$$

for some $m \times n$ matrix K (called *LQR gain matrix*)

- ▶ finding K involves taking correct 'slice' of inverse KKT matrix
- ▶ entries of K depend on horizon T , and converge as T grows large

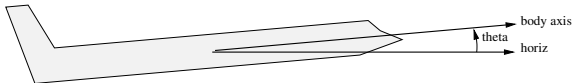
State feedback control

- ▶ find K for LQR problem (with large T)
- ▶ for each t ,
 - measure state x_t
 - implement control $u_t = Kx_t$
- ▶ with $u_t = Kx_t$ is called *state feedback control policy*
- ▶ combine with ('open-loop dynamics') $x_{t+1} = Ax_t + Bu_t$ to get *closed-loop dynamics*

$$x_{t+1} = (A + BK)x_t$$

- ▶ we can simulate open- and closed-loop dynamics to compare

Example: longitudinal flight control



variables are (small) deviations from operating point or *trim conditions*;

state is $x_t = (w_t, v_t, \theta_t, q_t)$:

- ▶ w_t : velocity of aircraft along body axis
- ▶ v_t : velocity of aircraft perpendicular to body axis (down is positive)
- ▶ θ_t : angle between body axis and horizontal (up is positive)
- ▶ $q_t = \dot{\theta}_t$: angular velocity of aircraft (pitch rate)

input is $u_t = (e_t, f_t)$:

- ▶ e_t : elevator angle ($e_t > 0$ is down)
- ▶ f_t : thrust

Linearized dynamics

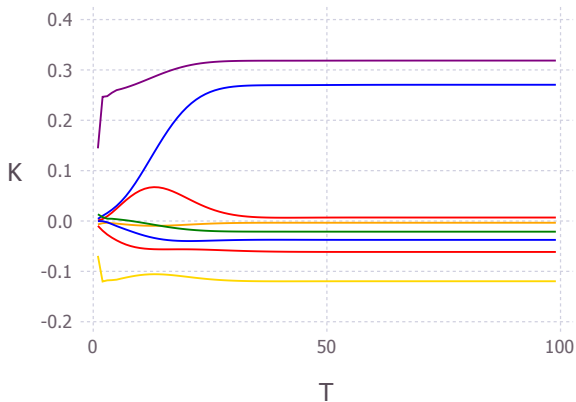
for 747, level flight, 40000 ft, 774 ft/sec, dynamics are
 $x_{t+1} = Ax_t + Bu_t$, where

$$A = \begin{bmatrix} .99 & .03 & -.02 & -.32 \\ .01 & .47 & 4.7 & .00 \\ .02 & -.06 & .40 & -.00 \\ .01 & -.04 & .72 & .99 \end{bmatrix}, \quad B = \begin{bmatrix} 0.01 & 0.99 \\ -3.44 & 1.66 \\ -0.83 & 0.44 \\ -0.47 & 0.25 \end{bmatrix}$$

- ▶ units: ft, sec, crad ($= 0.01\text{rad} \approx 0.57^\circ$)
- ▶ discretization is 1 sec

LQR gain

gain matrix K converged for $T \approx 30$



LQR for 747 model

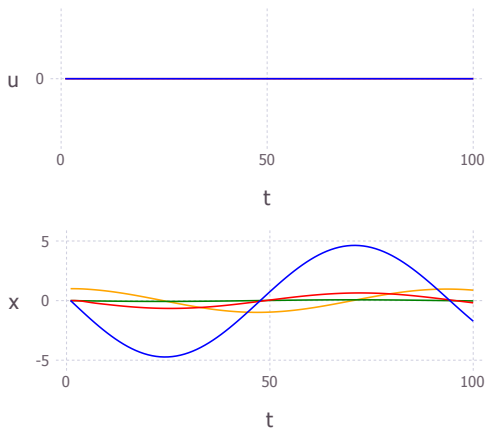
- ▶ LQR gain matrix (for $T = 100$, $\rho = 100$) is:

$$K = \begin{bmatrix} -.038 & .021 & .319 & -.270 \\ -.061 & -.004 & -.120 & .007 \end{bmatrix}$$

- ▶ e.g., $K_{14} = -.27$ is gain from pitch rate $((x_t)_4)$ to elevator angle $((u_t)_1)$

747 simulation

$u_t = 0$ ('open loop')



747 simulation

$$u_t = Kx_t \text{ ('closed loop')}$$

