Linear Systems

Linear Quadratic Optimal Control

Motivation

state feedback control:

- allows to arbitrarily assign the closed-loop eigenvalues for a controllable system
- the eigenvalue assignment has been manual thus far
- performance is implicit: we assign eigenvalues to induce proper error convergence

Motivation

state feedback control:

- allows to arbitrarily assign the closed-loop eigenvalues for a controllable system
- the eigenvalue assignment has been manual thus far
- performance is implicit: we assign eigenvalues to induce proper error convergence

linear quadratic (LQ) optimal regulation control, aka, LQ regulator (or LQR):

- no need to specify closed-loop poles
- performance is explicit: a performance index is defined ahead of time

1. Problem formulation

2. Solution to the infinite-horizon/stationary LQ problem

3. Solution to the finite-horizon LQ problem

4. From finite-horizon LQ to stationary LQ

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t), \;\; x(t_0) = x_0 \\ y(t) &= Cx(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t) \right) dt$$

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax\,(t) + Bu\,(t)\,,\;\; x\,(t_0) = x_0 \\ y\,(t) &= Cx\,(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- $S \succeq 0$, $Q \succeq 0$, $R \succ 0$: for a nonnegative cost and well-posed problem

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t), \; x(t_0) = x_0 \\ y(t) &= Cx(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- $S \succeq 0$, $Q \succeq 0$, $R \succ 0$: for a nonnegative cost and well-posed problem
- $\frac{1}{2}x^T(t_f)Sx(t_f)$ penalizes the deviation of x from the origin at t_f

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t), \; x(t_0) = x_0 \\ y(t) &= Cx(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2} x^T(t_f) S x(t_f) + \frac{1}{2} \int_{t_0}^{t_f} \left(x^T(t) Q x(t) + u^T(t) R u(t) \right) dt$$

- $S \succeq 0$, $Q \succeq 0$, $R \succ 0$: for a nonnegative cost and well-posed problem
- $\frac{1}{2} x^T(t_f) S x(t_f)$ penalizes the deviation of x from the origin at t_f
- $x^T(t) Q x(t)$ $t \in (t_0, t_f)$ penalizes the transient

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t), \ x(t_0) = x_0 \\ y(t) &= Cx(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- $S \succeq 0$, $Q \succeq 0$, $R \succ 0$: for a nonnegative cost and well-posed problem
- $\frac{1}{2}x^T(t_f)Sx(t_f)$ penalizes the deviation of x from the origin at t_f
- $x^T(t)Qx(t)$ $t \in (t_0, t_f)$ penalizes the transient
- often, $Q = C^T C \Rightarrow x^T(t)Qx(t) = y(t)^T y(t)$

Goal

Consider an n-dimensional state-space system

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t), \; x(t_0) = x_0 \\ y(t) &= Cx(t) \end{aligned} \tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and $y \in \mathbb{R}^m$.
LQ optimal control aims at minimizing the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- $S \succeq 0$, $Q \succeq 0$, $R \succ 0$: for a nonnegative cost and well-posed problem
- $\frac{1}{2}x^T(t_f)Sx(t_f)$ penalizes the deviation of x from the origin at t_f
- $x^T(t)Qx(t)$ $t \in (t_0, t_f)$ penalizes the transient
- often, $Q = C^TC \Rightarrow x^T(t)Qx(t) = y(t)^T y(t)$
- $u^T(t)Ru(t)$ penalizes large control efforts

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- when the control horizon is made to be infinitely long, i.e., $t_f \to \infty$, the problem reduces to the infinite-horizon LQ problem

$$J = \frac{1}{2}\int_{t_0}^{\infty} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f}\left(x^T(t)Qx(t) + u^T(t)Ru(t)\right)dt$$

- when the control horizon is made to be infinitely long, i.e., $t_f \to \infty$, the problem reduces to the infinite-horizon LQ problem

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x^T(t)Qx(t) + u^T(t)Ru(t)\right)dt$$

  - terminal cost is not needed, as it will turn out, that the control will have to drive x to the origin. Otherwise J will go unbounded.

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

- when the control horizon is made to be infinitely long, i.e., $t_f \to \infty$, the problem reduces to the infinite-horizon LQ problem

$$J = \frac{1}{2}\int_{t_0}^{\infty} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

  - terminal cost is not needed, as it will turn out, that the control will have to drive x to the origin. Otherwise J will go unbounded.
  - often, we have

$$J = \frac{1}{2}\int_{0}^{\infty} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

1. Problem formulation

2. Solution to the infinite-horizon/stationary LQ problem

3. Solution to the finite-horizon LQ problem

4. From finite-horizon LQ to stationary LQ

Solution concept: infinite-horizon/stationary LQ

Consider the performance index

$$J = \frac{1}{2} \int_{t_0}^{\infty} \left(x(t)^T Q x(t) + u(t)^T R u(t) \right) dt, \;\; Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t), \; x(t_0) = x_0$ and $R \succ 0$.

Solution concept: infinite-horizon/stationary LQ

Consider the performance index

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right) dt,\ \ Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t)$, $x(t_0) = x_0$ and $R \succ 0$.

- recall when addressing $\overline{J} = \frac{1}{2}\int_0^{\infty} x^T(t)\,Qx(t)\,dt,\ \ \dot{x} = Ax$

Solution concept: infinite-horizon/stationary LQ

Consider the performance index

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right)dt,\;\; Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t),\;\; x(t_0) = x_0$ and $R \succ 0$.

- recall when addressing $\overline{J} = \frac{1}{2}\int_0^\infty x^T(t)\,Qx(t)\,dt,\;\; \dot{x} = Ax$
- we defined $V(t) = \frac{1}{2}x^T(t)\,Px(t),\;\; P = P^T$, such that

$$\begin{aligned}\overline{J} + V(\infty) - V(0) &= \frac{1}{2}\int_0^\infty x^T(t)\,Qx(t)\,dt + \int_0^\infty \dot{V}(t)\,dt \\ &= \frac{1}{2}\int_0^\infty x^T(t)\left(Q + A^T P + PA\right)x(t)\,dt\end{aligned}$$

Solution concept: infinite-horizon/stationary LQ

Consider the performance index

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right)dt,\ \ Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t)\,,\ x(t_0) = x_0$ and $R \succ 0$.

- recall when addressing $\overline{J} = \frac{1}{2}\int_0^{\infty} x^T(t)\,Qx(t)\,dt,\ \dot{x} = Ax$
- we defined $V(t) = \frac{1}{2}x^T(t)\,Px(t)\,,\ P = P^T$, such that

$$\begin{aligned}\overline{J} + V(\infty) - V(0) &= \frac{1}{2}\int_0^{\infty} x^T(t)\,Qx(t)\,dt + \int_0^{\infty}\dot{V}(t)\,dt\\ &= \frac{1}{2}\int_0^{\infty} x^T(t)\left(Q + A^T P + PA\right)x(t)\,dt\end{aligned}$$

- yielding $\overline{J}^0 = \frac{1}{2}x^T(0)\,P_+ x(0)$ where P_+ comes from $A^T P + PA + Q = 0$, when the origin is asymptotically stable.

Solution of the infinite-horizon LQ

It turns out (see details in course notes) that for

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right)dt,\ \ Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t),\ x(t_0) = x_0$ and $R \succ 0$:

- if (A, B) is **controllable** (stabilizable) and (A, C) is **observable** (detectable)

Solution of the infinite-horizon LQ

It turns out (see details in course notes) that for

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right)dt,\ \ Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t),\ \ x(t_0) = x_0$ and $R \succ 0$:

- if (A, B) is **controllable** (stabilizable) and (A, C) is **observable** (detectable)
- then the optimal control input is given by

$$u(t) = -R^{-1}B^T P_+ x(t)$$

Solution of the infinite-horizon LQ

It turns out (see details in course notes) that for

$$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x(t)^T Qx(t) + u(t)^T Ru(t)\right) dt, \;\; Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t)\,, \;\; x(t_0) = x_0$ and $R \succ 0$:

- if (A, B) is **controllable** (stabilizable) and (A, C) is **observable** (detectable)
- then the optimal control input is given by

$$u(t) = -R^{-1}B^T P_+ x(t)$$

- where $P_+ \; (= P_+^T)$ is the positive (semi)definite solution of the **algebraic Riccati equation** (ARE)

$$A^T P + PA - PBR^{-1}B^T P + Q = 0$$

Solution of the infinite-horizon LQ

It turns out (see details in course notes) that for

$$J = \frac{1}{2}\int_{t_0}^{\infty} \left(x(t)^T Qx(t) + u(t)^T Ru(t)\right) dt, \;\; Q = C^T C$$

with $\dot{x}(t) = Ax(t) + Bu(t), \;\; x(t_0) = x_0$ and $R \succ 0$:

- if (A, B) is **controllable** (stabilizable) and (A, C) is **observable** (detectable)
- then the optimal control input is given by

$$u(t) = -R^{-1}B^T P_+ x(t)$$

- where $P_+ \left(= P_+^T\right)$ is the positive (semi)definite solution of the **algebraic Riccati equation** (ARE)

$$A^T P + PA - PBR^{-1}B^T P + Q = 0$$

- and the closed-loop system is **asymptotically stable**, with

$$J_{\min} = J^0 = \frac{1}{2}x(t_0)^T P_+ x(t_0)$$

Observations

- the control $u(t) = -R^{-1}B^T Px(t)$ is a state feedback law

Observations

- the control $u(t) = -R^{-1}B^T Px(t)$ is a state feedback law
- under the optimal control, the closed loop is given by $\dot{x} = Ax - BR^{-1}B^T Px = \underbrace{\left(A - BR^{-1}B^T P\right)}_{A_c} x$ and $J = \frac{1}{2}\int_{t_0}^{\infty} \left(x^T Qx + u^T Ru\right) dt = \frac{1}{2}\int_{t_0}^{\infty} x^T \underbrace{\left(Q + PBR^{-1}B^T P\right)}_{Q_c} xdt$

Observations

- the control $u(t) = -R^{-1}B^TPx(t)$ is a state feedback law
- under the optimal control, the closed loop is given by $\dot{x} = Ax - BR^{-1}B^TPx = \underbrace{\left(A - BR^{-1}B^TP\right)}_{A_c} x$ and $J = \frac{1}{2}\int_{t_0}^{\infty}\left(x^TQx + u^TRu\right)dt = \frac{1}{2}\int_{t_0}^{\infty} x^T \underbrace{\left(Q + PBR^{-1}B^TP\right)}_{Q_c} xdt$
- for the above closed-loop system, the Lyapunov Eq. is

$$
\begin{aligned}
A_c^TP + PA_c &= -Q_c \\
\Leftrightarrow \left(A - BR^{-1}B^TP\right)^T P + P\left(A - BR^{-1}B^TP\right) &= -Q - PBR^{-1}B^TP \\
\Leftrightarrow A^TP + PA - PBR^{-1}B^TP &= -Q \Leftarrow \text{the ARE!}
\end{aligned}
$$

Observations

- the control $u(t) = -R^{-1}B^T Px(t)$ is a state feedback law
- under the optimal control, the closed loop is given by $\dot{x} = Ax - BR^{-1}B^T Px = \underbrace{\left(A - BR^{-1}B^T P\right)}_{A_c} x$ and $J =$
$$\frac{1}{2}\int_{t_0}^{\infty}\left(x^T Qx + u^T Ru\right)dt = \frac{1}{2}\int_{t_0}^{\infty} x^T \underbrace{\left(Q + PBR^{-1}B^T P\right)}_{Q_c} xdt$$
- for the above closed-loop system, the Lyapunov Eq. is
$$\begin{aligned} A_c^T P + PA_c &= -Q_c \\ \Leftrightarrow \left(A - BR^{-1}B^T P\right)^T P + P\left(A - BR^{-1}B^T P\right) &= -Q - PBR^{-1}B^T P \\ \Leftrightarrow A^T P + PA - PBR^{-1}B^T P &= -Q \Leftarrow \text{the ARE!} \end{aligned}$$
- when the ARE solution P_+ is positive definite, $\frac{1}{2}x^T P_+ x$ is a Lyapunov function for the closed-loop system

Observations

- Lyapunov Eq. and the ARE:

Cost	$\overline{J} = \frac{1}{2}\int_0^\infty x^T Q x dt$	$J = \frac{1}{2}\int_{t_0}^\infty \left(x^T Q x + u^T R u\right) dt$
Syst. dynamics	$\dot{x} = Ax$	$\dot{x} = Ax + Bu$ (A, B) controllable/stabilizable (A, C) observable/detectable
Key Eq.	$A^T P + PA + Q = 0$	$A^T P + PA - PBR^{-1}B^T P + Q = 0$
Optimal control	N/A	$u(t) = -R^{-1}B^T P_+ x(t)$
Opt. cost	$\overline{J}^0 = \frac{1}{2}x^T(0)\, P_+ x(0)$	$J^0 = \frac{1}{2}x(t_0)^T\, P_+ x(t_0)$

Observations

- Lyapunov Eq. and the ARE:

Cost	$\overline{J} = \frac{1}{2}\int_0^\infty x^T Q x dt$	$J = \frac{1}{2}\int_{t_0}^\infty \left(x^T Q x + u^T R u\right) dt$
Syst. dynamics	$\dot{x} = Ax$	$\dot{x} = Ax + Bu$ (A, B) controllable/stabilizable (A, C) observable/detectable
Key Eq.	$A^T P + PA + Q = 0$	$A^T P + PA - PBR^{-1}B^T P + Q = 0$
Optimal control	N/A	$u(t) = -R^{-1}B^T P_+ x(t)$
Opt. cost	$\overline{J}^0 = \frac{1}{2}x^T(0)\, P_+ x(0)$	$J^0 = \frac{1}{2}x(t_0)^T\, P_+ x(t_0)$

- the guaranteed closed-loop stability is an attractive feature
- more nice properties will show up later

Example: Stationary LQR of a pure inertia system

- Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \; J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt, \; R > 0$$

Example: Stationary LQR of a pure inertia system

- Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u,\ J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt,\ R > 0$$

- the ARE is

$$0 = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} P + P \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - P \begin{bmatrix} 0 \\ 1 \end{bmatrix} \frac{1}{R} \begin{bmatrix} 0 & 1 \end{bmatrix} P \Rightarrow P_+ = \begin{bmatrix} \sqrt{2} R^{1/4} & R^{1/2} \\ R^{1/2} & \sqrt{2} R^{3/4} \end{bmatrix}$$

Example: Stationary LQR of a pure inertia system

- Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \; J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt, \; R > 0$$

- the ARE is

$$0 = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} P + P \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - P \begin{bmatrix} 0 \\ 1 \end{bmatrix} \frac{1}{R} \begin{bmatrix} 0 & 1 \end{bmatrix} P \Rightarrow P_+ = \begin{bmatrix} \sqrt{2}R^{1/4} & R^{1/2} \\ R^{1/2} & \sqrt{2}R^{3/4} \end{bmatrix}$$

- the closed-loop A matrix can be computed to be

$$A_c = A - BR^{-1}B^T P_+ = \begin{bmatrix} 0 & 1 \\ -R^{-1/2} & -\sqrt{2}R^{-1/4} \end{bmatrix}$$

Example: Stationary LQR of a pure inertia system

- Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \; J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt, \; R > 0$$

- the ARE is

$$0 = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} P + P \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - P \begin{bmatrix} 0 \\ 1 \end{bmatrix} \frac{1}{R} \begin{bmatrix} 0 & 1 \end{bmatrix} P \Rightarrow P_+ = \begin{bmatrix} \sqrt{2}R^{1/4} & R^{1/2} \\ R^{1/2} & \sqrt{2}R^{3/4} \end{bmatrix}$$

- the closed-loop A matrix can be computed to be

$$A_c = A - BR^{-1}B^T P_+ = \begin{bmatrix} 0 & 1 \\ -R^{-1/2} & -\sqrt{2}R^{-1/4} \end{bmatrix}$$

- $\Rightarrow$ closed-loop eigenvalues:

$$\lambda_{1,2} = -\frac{1}{\sqrt{2}R^{1/4}} \pm \frac{1}{\sqrt{2}R^{1/4}} j$$

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \;\; J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt$$

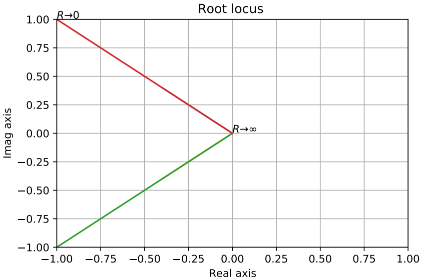


Figure: Eigenvalue $\lambda_{1,2} = -\frac{1}{\sqrt{2}R^{1/4}} \pm \frac{1}{\sqrt{2}R^{1/4}} j$ evolution (root locus)

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \;\; J = \frac{1}{2}\int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt$$

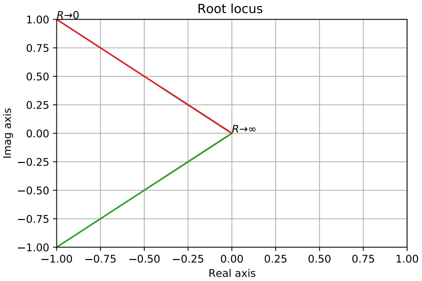


Figure: Eigenvalue $\lambda_{1,2} = -\frac{1}{\sqrt{2}R^{1/4}} \pm \frac{1}{\sqrt{2}R^{1/4}} j$ evolution (root locus)

- $R \uparrow$ (more penalty on the control input) $\Rightarrow \lambda_{1,2}$ move closer to the origin $\Rightarrow$ slower state convergence to zero

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \;\; J = \frac{1}{2} \int_0^\infty \left(x^T \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} x + Ru^2 \right) dt$$

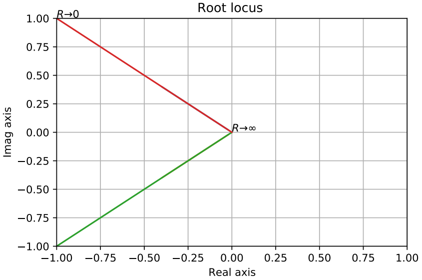


Figure: Eigenvalue $\lambda_{1,2} = -\frac{1}{\sqrt{2}R^{1/4}} \pm \frac{1}{\sqrt{2}R^{1/4}} j$ evolution (root locus)

- $R \uparrow$ (more penalty on the control input) $\Rightarrow \lambda_{1,2}$ move closer to the origin $\Rightarrow$ slower state convergence to zero
- $R \downarrow$ (allow for large control efforts) $\Rightarrow \lambda_{1,2}$ move further to the left of the complex plane $\Rightarrow$ faster speed of closed-loop dynamics

MATLAB commands

- *care*: solves the ARE for a continuous-time system:

$$[P, \Lambda, K] = \text{care}\left(A, B, C^T C, R\right)$$

where $K = R^{-1}B^T P$ and Λ is a diagonal matrix with the closed-loop eigenvalues, i.e., the eigenvalues of $A - BK$, in the diagonal entries.

MATLAB commands

- *care*: solves the ARE for a continuous-time system:

$$[P, \Lambda, K] = \text{care}\left(A, B, C^T C, R\right)$$

where $K = R^{-1}B^T P$ and Λ is a diagonal matrix with the closed-loop eigenvalues, i.e., the eigenvalues of $A - BK$, in the diagonal entries.

- *lqr* and *lqry*: provide the LQ regulator with

$$[K, P, \Lambda] = \text{lqr}\left(A, B, C^T C, R\right)$$
$$[K, P, \Lambda] = \text{lqry}\left(\text{sys}, Q_y, R\right)$$

where *sys* is defined by $\dot{x} = Ax + Bu, \;\; y = Cx + Du$, and

$$J = \frac{1}{2}\int_0^{\infty} \left(y^T Q_y y + u^T R u\right) dt$$

Solution to the finite-horizon LQ

Consider the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

with $\dot{x} = Ax + Bu,\ x\,(t_0) = x_0$, $S \succeq 0$, $R \succ 0$, and $Q = C^T C$.

Solution to the finite-horizon LQ

Consider the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

with $\dot{x} = Ax + Bu,\ x(t_0) = x_0$, $S \succeq 0$, $R \succ 0$, and $Q = C^T C$.

- do a similar Lyapunov construction: $V(t) \triangleq \frac{1}{2}x^T(t)\,P(t)\,x(t)$

Solution to the finite-horizon LQ

Consider the performance index

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

with $\dot{x} = Ax + Bu,\ x(t_0) = x_0$, $S \succeq 0$, $R \succ 0$, and $Q = C^T C$.

- do a similar Lyapunov construction: $V(t) \triangleq \frac{1}{2}x^T(t)\,P(t)\,x(t)$
- then

$$\begin{aligned}
\frac{d}{dt}V(t) &= \frac{1}{2}\dot{x}^T(t)\,P(t)\,x(t) + \frac{1}{2}x^T(t)\,\dot{P}(t)\,x(t) + \frac{1}{2}x^T(t)\,P(t)\,\dot{x}(t) \\
&= \frac{1}{2}(Ax+Bu)^T Px + \frac{1}{2}x^T\frac{dP}{dt}x + \frac{1}{2}x^T P(Ax+Bu) \\
&= \frac{1}{2}\left\{x^T(t)\left(A^T P + \frac{dP}{dt} + PA\right)x(t) + u^T B^T Px + x^T PBu\right\}
\end{aligned}$$

Solution to the finite-horizon LQ

with $\frac{d}{dt}V(t)$ from the last slide, we have

$$
\begin{aligned}
V(t_f) - V(t_0) &= \int_{t_0}^{t_f} \dot{V}\,dt \\
&= \frac{1}{2}\int_{t_0}^{t_f} \left(x^T \left(A^T P + PA + \frac{dP}{dt} \right) x + u^T B^T P x + x^T P B u \right) dt
\end{aligned}
$$

Solution to the finite-horizon LQ

with $\frac{d}{dt}V(t)$ from the last slide, we have

$$V(t_f) - V(t_0) = \int_{t_0}^{t_f} \dot{V}\, dt$$
$$= \frac{1}{2}\int_{t_0}^{t_f} \left(x^T\left(A^TP + PA + \frac{dP}{dt}\right)x + u^TB^TPx + x^TPBu\right) dt$$

- adding

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

  to both sides yields

$$J + V(t_f) - V(t_0) = \frac{1}{2}x^T(t_f)Sx(t_f) +$$
$$\frac{1}{2}\int_{t_0}^{t_f} \left(x^T\left(A^TP + PA + Q + \frac{dP}{dt}\right)x + \underbrace{u^TB^TPx + x^TPBu}_{\text{products of } x \text{ and } u} + \underbrace{u^TRu}_{\text{quadratic}}\right) dt$$

Solution to the finite-horizon LQ

- "complete the squares" in $\underbrace{u^T B^T P x + x^T P B u}_{\text{products of } x \text{ and } u} + \underbrace{u^T R u}_{\text{quadratic}}$:

$$
\begin{aligned}
& u^T B^T P x + x^T P B u + u^T R u \overset{\text{scalar case}}{=} R u^2 + 2 u B P x \\
=& R u^2 + 2 \left(x P B R^{-1/2} \right) \underbrace{R^{1/2} u}_{\sqrt{R u^2}} + \left(R^{-1/2} B P x \right)^2 - \left(R^{-1/2} B P x \right)^2 \\
=& \left(R^{1/2} u + R^{-1/2} B P x \right)^2 - \left(R^{-1/2} B P x \right)^2
\end{aligned}
$$

Solution to the finite-horizon LQ

- "complete the squares" in $\underbrace{u^T B^T P x + x^T P B u}_{\text{products of } x \text{ and } u} + \underbrace{u^T R u}_{\text{quadratic}}$:

$$u^T B^T P x + x^T P B u + u^T R u \overset{\text{scalar case}}{=} R u^2 + 2uBPx$$

$$= R u^2 + 2\left(xPBR^{-1/2}\right)\underbrace{R^{1/2}u}_{\sqrt{Ru^2}} + \left(R^{-1/2}BPx\right)^2 - \left(R^{-1/2}BPx\right)^2$$

$$= \left(R^{1/2}u + R^{-1/2}BPx\right)^2 - \left(R^{-1/2}BPx\right)^2$$

- extending the concept to the general vector case:

$$u^T B^T P x + x^T P B u + u^T R u = \|R^{\frac{1}{2}} u + R^{\frac{-1}{2}} B^T P x\|_2^2 - x^T P B R^{-1} B^T P x$$

Solution to the finite-horizon LQ

$$J + V(t_f) - V(t_0) = \frac{1}{2}x^T(t_f)Sx(t_f) +$$
$$\frac{1}{2}\int_{t_0}^{t_f}\left(x^T\left(A^TP + PA + Q + \frac{dP}{dt}\right)x + u^TB^TPx + x^TPBu + u^TRu\right)dt$$

$\Downarrow$"completing the squares"

$$J + \underline{\frac{1}{2}x^T(t_f)P(t_f)x(t_f)} - \frac{1}{2}x^T(t_0)P(t_0)x(t_0) = \underline{\frac{1}{2}x^T(t_f)Sx(t_f)} +$$
$$\frac{1}{2}\int_{t_0}^{t_f}\left(x^T\underline{\underline{\left(\frac{dP}{dt} + A^TP + PA + Q - PBR^{-1}B^TP\right)}}x + \|\underline{\underline{\underline{R^{\frac{1}{2}}u + R^{\frac{-1}{2}}B^TPx}}}\|_2^2\right)dt$$

Solution to the finite-horizon LQ

$$J + V(t_f) - V(t_0) = \frac{1}{2}x^T(t_f)Sx(t_f) +$$
$$\frac{1}{2}\int_{t_0}^{t_f}\left(x^T\left(A^TP + PA + Q + \frac{dP}{dt}\right)x + u^TB^TPx + x^TPBu + u^TRu\right)dt$$

$\Downarrow$"completing the squares"

$$J + \underline{\frac{1}{2}x^T(t_f)P(t_f)x(t_f)} - \frac{1}{2}x^T(t_0)P(t_0)x(t_0) = \underline{\frac{1}{2}x^T(t_f)Sx(t_f)} +$$
$$\frac{1}{2}\int_{t_0}^{t_f}\left(x^T\underline{\underline{\left(\frac{dP}{dt} + A^TP + PA + Q - PBR^{-1}B^TP\right)}}x + \|\underline{\underline{\underline{R^{\frac{1}{2}}u + R^{\frac{-1}{2}}B^TPx}}}\|_2^2\right)dt$$

- the best that the control can do in minimizing the cost is to have

$$u(t)\underline{\underline{\underline{= -K(t)x(t) = -R^{-1}B^TP(t)x(t)}}}$$
$$-\frac{dP}{dt}\underline{\underline{= A^TP + PA - PBR^{-1}B^TP + Q}},\ \underline{P(t_f) = S}$$

to yield the optimal cost $J^0 = \frac{1}{2}x_0^TP(t_0)x_0$

Observations

$$u(t) = -K(t)\,x(t) = -R^{-1}B^T P(t)x(t) \qquad \text{optimal state feedback control}$$

$$-\frac{dP}{dt} = A^T P + PA - PBR^{-1}B^T P + Q,\ \ P(t_f) = S \qquad \text{the Riccati differential equation}$$

- boundary condition of the Riccati equation is given at the final time $t_f \Rightarrow$ the equation must be integrated backward in time

Observations

$$u(t) = -K(t)\,x(t) = -R^{-1}B^{T}P(t)x(t) \qquad \text{optimal state feedback control}$$

$$-\frac{dP}{dt} = A^{T}P + PA - PBR^{-1}B^{T}P + Q,\ P(t_f) = S \qquad \text{the Riccati differential equation}$$

- boundary condition of the Riccati equation is given at the final time $t_f \Rightarrow$ the equation must be integrated backward in time
- *backward* integration of

  $$-\frac{dP}{dt} = A^{T}P + PA + Q - PBR^{-1}B^{T}P,\ P(t_f) = S$$

  is equivalent to the *forward* integration of

  $$\frac{dP^*}{dt} = A^{T}P^* + P^*A + Q - P^*BR^{-1}B^{T}P^*,\ P^*(0) = S \qquad (2)$$

  by letting $P(t) = P^*(t_f - t)$

Observations

$$u(t) = -K(t)\,x(t) = -R^{-1}B^T P(t)x(t) \qquad \text{optimal state feedback control}$$

$$-\frac{dP}{dt} = A^T P + PA - PBR^{-1}B^T P + Q,\ P(t_f) = S \qquad \text{the Riccati differential equation}$$

- boundary condition of the Riccati equation is given at the final time $t_f \Rightarrow$ the equation must be integrated backward in time
- *backward* integration of

  $$-\frac{dP}{dt} = A^T P + PA + Q - PBR^{-1}B^T P,\ P(t_f) = S$$

  is equivalent to the *forward* integration of

  $$\frac{dP^*}{dt} = A^T P^* + P^* A + Q - P^* BR^{-1}B^T P^*,\ P^*(0) = S \qquad (2)$$

  by letting $P(t) = P^*(t_f - t)$
- Eq. (2) can be solved by numerical integration, e.g., ODE45 in Matlab

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

$$J^0 = \frac{1}{2}x_0^T P(t_0)x_0$$

- the minimum value J^0 is a function of the initial state $x(t_0)$

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

$$J^0 = \frac{1}{2}x_0^T P(t_0)x_0$$

- the minimum value J^0 is a function of the initial state $x(t_0)$
- J (and hence J^0) is nonnegative $\Rightarrow$ $P(t_0)$ is at least positive semidefinite

Observations

$$J = \frac{1}{2}x^T(t_f)Sx(t_f) + \frac{1}{2}\int_{t_0}^{t_f} \left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$$

$$J^0 = \frac{1}{2}x_0^T P(t_0)x_0$$

- the minimum value J^0 is a function of the initial state $x(t_0)$
- J (and hence J^0) is nonnegative $\Rightarrow P(t_0)$ is at least positive semidefinite
- t_0 can be taken anywhere in $(0, t_f) \Rightarrow P(t)$ is at least positive semidefinite for any t

Observations

$$J = \frac{1}{2} x^T(t_f) S x(t_f) + \frac{1}{2} \int_{t_0}^{t_f} \left(x^T(t) Q x(t) + u^T(t) R u(t) \right) dt$$

$$J^0 = \frac{1}{2} x_0^T P(t_0) x_0$$

- the minimum value J^0 is a function of the initial state $x(t_0)$
- J (and hence J^0) is nonnegative $\Rightarrow P(t_0)$ is at least positive semidefinite
- t_0 can be taken anywhere in $(0, t_f) \Rightarrow P(t)$ is at least positive semidefinite for any t
- the state feedback law is time varying because of $P(t)$

Example: LQR of a pure inertia system

Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \;\; J = \frac{1}{2} x^T(t_f) S x(t_f) + \frac{1}{2} \int_0^{t_f} \left(x^T Q x + R u^2 \right) dt$$

where $S = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \;\; Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \;\; R > 0$

Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \; J = \frac{1}{2} x^T(t_f) S x(t_f) + \frac{1}{2} \int_0^{t_f} \left(x^T Q x + R u^2 \right) dt$$

where $S = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \; Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \; R > 0$

- we let $P(t) = P^*(t_f - t)$ and solve

$$\frac{dP^*}{dt} = A^T P^* + P^* A + Q - P^* B R^{-1} B^T P^*, \; P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Leftrightarrow \frac{dP^*}{dt} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} P^* + P^* \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - P^* \begin{bmatrix} 0 \\ 1 \end{bmatrix} \frac{1}{R} \begin{bmatrix} 0 & 1 \end{bmatrix} P^*$$

Example: LQR of a pure inertia system

Consider

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \;\; J = \frac{1}{2} x^T(t_f) S x(t_f) + \frac{1}{2}\int_0^{t_f} \left(x^T Q x + R u^2 \right) dt$$

where $S = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \;\; Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \;\; R > 0$

- we let $P(t) = P^*(t_f - t)$ and solve

$$\frac{dP^*}{dt} = A^T P^* + P^* A + Q - P^* B R^{-1} B^T P^*, \;\; P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Leftrightarrow \frac{dP^*}{dt} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} P^* + P^* \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - P^* \begin{bmatrix} 0 \\ 1 \end{bmatrix} \frac{1}{R} \begin{bmatrix} 0 & 1 \end{bmatrix} P^*$$

- letting

$$P^* = \begin{bmatrix} p_{11}^* & p_{12}^* \\ p_{12}^* & p_{22}^* \end{bmatrix} \Rightarrow \begin{cases} \frac{d}{dt} p_{11}^* = 1 - \frac{1}{R}\left(p_{12}^*\right)^2 & p_{11}^*(0) = 1 \\ \frac{d}{dt} p_{12}^* = p_{11}^* - \frac{1}{R} p_{12}^* p_{22}^* & p_{12}^*(0) = 0 \\ \frac{d}{dt} p_{22}^* = 2 p_{12}^* - \frac{1}{R}\left(p_{22}^*\right)^2 & p_{22}^*(0) = 1 \end{cases}$$

Example: LQR of a pure inertia system: analysis

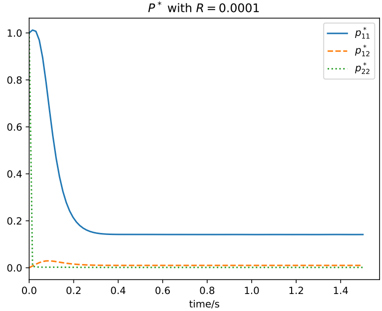


Figure: LQ example: $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $P(t) = P^*(t_f - t)$

Example: LQR of a pure inertia system: analysis

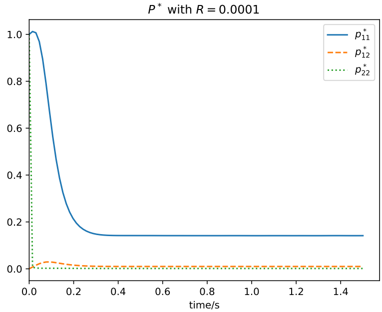


Figure: LQ example: $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $P(t) = P^*(t_f - t)$

- if the final time t_f is large, $P^*(t)$ forward converges to a stationary value

Example: LQR of a pure inertia system: analysis

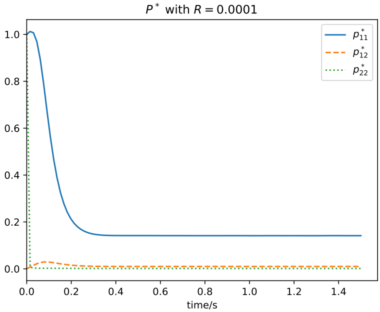


Figure: LQ example: $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $P(t) = P^*(t_f - t)$

- if the final time t_f is large, $P^*(t)$ forward converges to a stationary value
- i.e., $P(t)$ backward converges to a stationary value at $P(0)$

Example: LQR of a pure inertia system: analysis

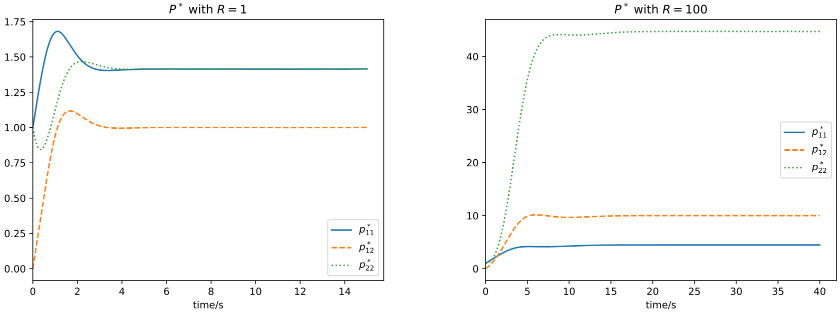
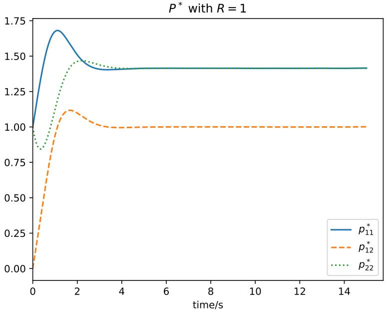


Figure: LQ example with different penalties on control. $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

Example: LQR of a pure inertia system: analysis



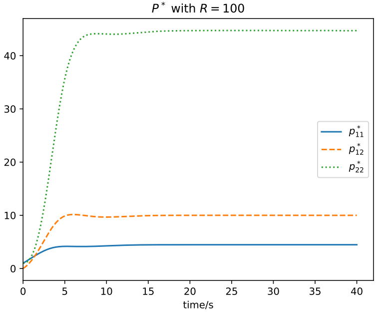


Figure: LQ example with different penalties on control. $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

- a larger R results in a longer transient

Example: LQR of a pure inertia system: analysis

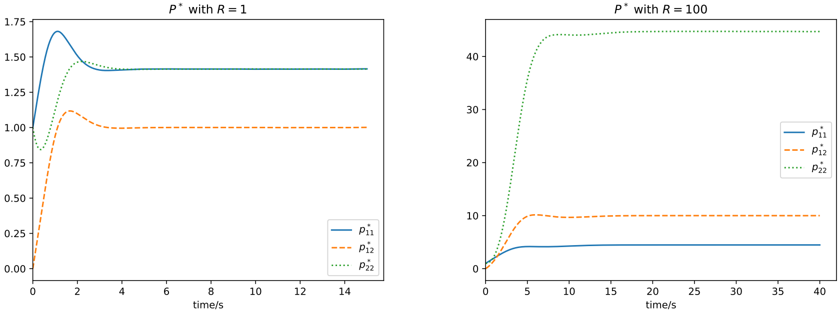
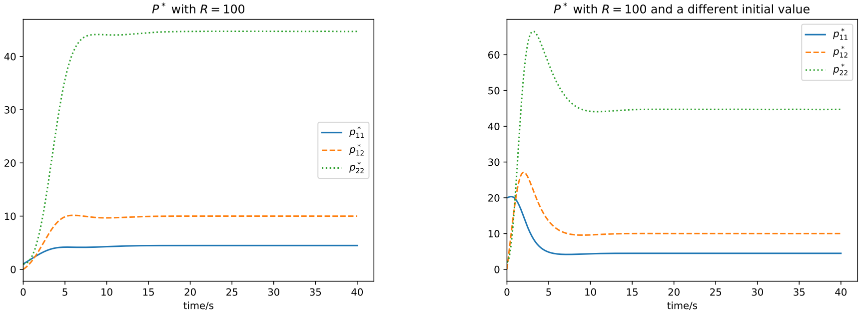


Figure: LQ example with different penalties on control. $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

- a larger R results in a longer transient
- i.e., a larger penalty on the control input yields a longer time to settle

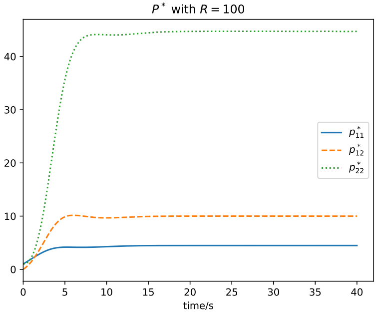
Example: LQR of a pure inertia system: analysis



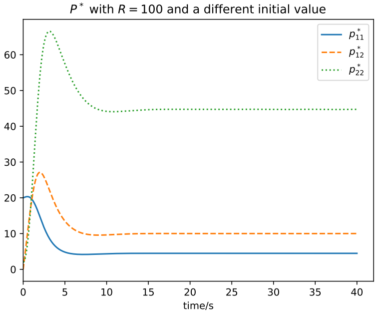
(a) $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ (b) $P^*(0) = \begin{bmatrix} 20 & 0 \\ 0 & 2 \end{bmatrix}$

Figure: LQ with different boundary values in Riccati difference Eq.

Example: LQR of a pure inertia system: analysis



(a) $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$



(b) $P^*(0) = \begin{bmatrix} 20 & 0 \\ 0 & 2 \end{bmatrix}$

Figure: LQ with different boundary values in Riccati difference Eq.

- for the same R, the initial value $P(t_f) = S$ becomes irrelevant

1. Problem formulation

2. Solution to the infinite-horizon/stationary LQ problem

3. Solution to the finite-horizon LQ problem

4. From finite-horizon LQ to stationary LQ

From LQ to stationary LQ

- the ARE and the Riccati differential Eq.:

Cost	$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x^T Q x + u^T R u\right) dt$	$J = \frac{1}{2}x^T(t_f)Sx(t_f) +$ $\frac{1}{2}\int_{t_0}^{t_f}\left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$
Syst.	$\dot{x} = Ax + Bu$ (A, B) controllable/stabilizable (A, C) observable/detectable	$\dot{x} = Ax + Bu$
Key Eq.	$A^T P + PA - PBR^{-1}B^T P + Q = 0$	$-\frac{dP}{dt} = A^T P + PA - PBR^{-1}B^T P + Q$ $P(t_f) = S$
Opt. control	$u(t) = -R^{-1}B^T P_+ x(t)$	$u(t) = -R^{-1}B^T P(t)x(t)$
Opt. cost	$J^0 = \frac{1}{2}x_0^T P_+ x_0$	$J^0 = \frac{1}{2}x_0^T P(t_0)x_0$

From LQ to stationary LQ

- the ARE and the Riccati differential Eq.:

Cost	$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x^T Q x + u^T R u\right) dt$	$J = \frac{1}{2}x^T(t_f)Sx(t_f) +$ $\frac{1}{2}\int_{t_0}^{t_f}\left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$
Syst.	$\dot{x} = Ax + Bu$ (A, B) controllable/stabilizable (A, C) observable/detectable	$\dot{x} = Ax + Bu$
Key Eq.	$A^T P + PA - PBR^{-1}B^T P + Q = 0$	$-\frac{dP}{dt} = A^T P + PA - PBR^{-1}B^T P + Q$ $P(t_f) = S$
Opt. control	$u(t) = -R^{-1}B^T P_+ x(t)$	$u(t) = -R^{-1}B^T P(t)x(t)$
Opt. cost	$J^0 = \frac{1}{2}x_0^T P_+ x_0$	$J^0 = \frac{1}{2}x_0^T P(t_0)x_0$

- in the example, we see that P in the Riccati differential Eq. converges to a stationary value given sufficient time

From LQ to stationary LQ

- the ARE and the Riccati differential Eq.:

Cost	$J = \frac{1}{2}\int_{t_0}^{\infty}\left(x^T Q x + u^T R u\right) dt$	$J = \frac{1}{2}x^T(t_f)Sx(t_f) +$ $\frac{1}{2}\int_{t_0}^{t_f}\left(x^T(t)Qx(t) + u^T(t)Ru(t)\right) dt$
Syst.	$\dot{x} = Ax + Bu$ (A, B) controllable/stabilizable (A, C) observable/detectable	$\dot{x} = Ax + Bu$
Key Eq.	$A^T P + PA - PBR^{-1}B^T P + Q = 0$	$-\frac{dP}{dt} = A^T P + PA - PBR^{-1}B^T P + Q$ $P(t_f) = S$
Opt. control Opt. cost	$u(t) = -R^{-1}B^T P_+ x(t)$ $J^0 = \frac{1}{2}x_0^T P_+ x_0$	$u(t) = -R^{-1}B^T P(t)x(t)$ $J^0 = \frac{1}{2}x_0^T P(t_0)x_0$

- in the example, we see that P in the Riccati differential Eq. converges to a stationary value given sufficient time
- when $t_f \to \infty$, the Riccati differential Eq. converges to ARE and the LQ becomes the stationary LQ, under two conditions that we now discuss in details:
  - (A, B) is controllable/stabilizable
  - (A, C) is observable/detectable

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2} x_0^T P(0) x_0$ is unbounded, we will have $||P(0)|| = \infty$

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2} x_0^T P(0) x_0$ is unbounded, we will have $||P(0)|| = \infty$
- e.g.: $\dot{x} = x + 0 \cdot u$, $x(0) = 1$, $Q = 1$ and R be any positive value

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2} x_0^T P(0) x_0$ is unbounded, we will have $||P(0)|| = \infty$
- e.g.: $\dot{x} = x + 0 \cdot u$, $x(0) = 1$, $Q = 1$ and R be any positive value
  - system is uncontrollable and the uncontrollable mode is unstable

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2}x_0^T P(0) x_0$ is unbounded, we will have $||P(0)|| = \infty$
- e.g.: $\dot{x} = x + 0 \cdot u$, $x(0) = 1$, $Q = 1$ and R be any positive value
  - system is uncontrollable and the uncontrollable mode is unstable
  - $x(t)$ will keep increasing to infinity

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2} x_0^T P(0)\, x_0$ is unbounded, we will have $||P(0)|| = \infty$
- e.g.: $\dot{x} = x + 0 \cdot u$, $x(0) = 1$, $Q = 1$ and R be any positive value
  - system is uncontrollable and the uncontrollable mode is unstable
  - $x(t)$ will keep increasing to infinity
  - $\Rightarrow J = \frac{1}{2} \int_0^\infty \left(x^T Q x + u^T R u\right) dt$ unbounded regardless of $u(t)$

Need for controllability/stabilizability

if (A, B) is controllable or stabilizable, then $P(t)$ is guaranteed to converge to a bounded and stationary value

- for uncontrollable or unstabilizable systems, there can be unstable uncontrollable modes that cause J to be unbounded
- then if $J^0 = \frac{1}{2} x_0^T P(0) x_0$ is unbounded, we will have $||P(0)|| = \infty$
- e.g.: $\dot{x} = x + 0 \cdot u$, $x(0) = 1$, $Q = 1$ and R be any positive value
  - system is uncontrollable and the uncontrollable mode is unstable
  - $x(t)$ will keep increasing to infinity
  - $\Rightarrow J = \frac{1}{2} \int_0^\infty \left(x^T Q x + u^T R u\right) dt$ unbounded regardless of $u(t)$
  - in this case, the Riccati equation is

$$-\frac{dP}{dt} = P + P + 1 = 2P + 1 \Leftrightarrow \frac{dP^*}{dt} = 2P^* + 1$$

forward integration of P^* (backward integration of P), will drive $P^*(\infty)$ and $P(0)$ to infinity

Need for observability/detectability

if (A, C) is observable or detectable, the optimal state feedback control system will be asymptotically stable

Need for observability/detectability

if (A, C) is observable or detectable, the optimal state feedback control system will be asymptotically stable

- *intuition*: if the system is observable, $y = Cx$ will relate to all states $\Rightarrow$ regulating $x^T Q x = x^T C^T C x$ will regulate all states

Need for observability/detectability

if (A, C) is observable or detectable, the optimal state feedback control system will be asymptotically stable

- *intuition*: if the system is observable, $y = Cx$ will relate to all states $\Rightarrow$ regulating $x^T Q x = x^T C^T C x$ will regulate all states
- *formally*: if (A, C) is observable (detectable), the solution of the Riccati equation will converge to a positive (semi)definite value P_+ (proof in course notes)

Additional excellent properties of stationary LQ

- we know stationary LQR yields guaranteed closed-loop stability for controllable (stabilizable) and observable (detectable) systems

It turns out that LQ regulators with full state feedback has excellent additional properties of:

- at least a 60 degree phase margin
- infinite gain margin
- stability is guaranteed up to a 50% reduction in the gain

Applications and practice

choosing R and Q:

- if there is not a good idea for the structure for Q and R, start with diagonal matrices;

Applications and practice

choosing R and Q:

- if there is not a good idea for the structure for Q and R, start with diagonal matrices;
- gain an idea of the magnitude of each state variable and input variable
- call them $x_{i,\max}$ $(i = 1, \ldots, n)$ and $u_{i,\max}$ $(i = 1, \ldots, r)$
- make the diagonal elements of Q and R inversely proportional to $||x_{i,\max}||^2$ and $||u_{i,\max}||^2$, respectively.