ME 547: Linear Systems
Linear Quadratic Optimal Control

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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 finite-horizon LQ problem

3. 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

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

- $S \succeq 0, Q \succeq 0, R \succ 0$: 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

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

  - 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 $t_0 = 0$ and

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

1. Problem formulation

2. Solution to the finite-horizon LQ problem

3. From finite-horizon LQ to stationary LQ

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 Lyapunov-like 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

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

- adding

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

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 $\underbrace{u^T B^T P x + x^T P B u}_{\text{products of } x \text{ and } u} + \underbrace{u^T R u}_{\text{quadratic}}$ (scalar case):

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

- extending the concept to the general vector case:

$$
u^T B^T P x + x^T P B u + u^T R u = \underbrace{\|R^{\frac{1}{2}}u + R^{\frac{-1}{2}}B^T P x\|_2^2}_{\text{recall } \|\vec{a}\|_2^2 = \vec{a}^T \vec{a}} - x^T P B R^{-1} B^T P x
$$

Solution to the finite-horizon LQ

$$J + V(t_f) - V(t_0) = \frac{1}{2}x^T(t_f)Sx(t_f) +$$

$$\frac{1}{2}\int_{t_0}^{t_f} \left(x^T \left(A^T P + PA + Q + \frac{dP}{dt} \right) x + \underbrace{u^T B^T P x + x^T P B u + u^T R u}_{\|R^{\frac{1}{2}}u + R^{\frac{-1}{2}} B^T P x\|_2^2 - x^T P B R^{-1} B^T P x} \right) dt$$

⇓"completing the squares"

$$J + \frac{1}{2}x^T(t_f)P(t_f)x(t_f) - \frac{1}{2}x^T(t_0)P(t_0)x(t_0) = \frac{1}{2}x^T(t_f)Sx(t_f) +$$

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

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

$$u(t)=-K(t)x(t)=\underline{\underline{\underline{-R^{-1}B^TP(t)x(t)}}}$$

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

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

Observation 1

$$u(t) = -K(t)\,x(t) = -R^{-1}B^TP(t)x(t) \qquad \text{optimal control law}$$

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

- the control $u(t) = -R^{-1}B^TP(t)\,x(t)$ is a state feedback law (the power of state feedback!)
- the state feedback law is time-varying because of $P(t)$
- the closed-loop dynamics becomes

$$\dot{x}(t) = Ax(t) + Bu(t) = \underbrace{\left(A - BR^{-1}B^TP(t)\right)}_{\text{time-varying closed-loop dynamics}} x(t)$$

Observation 2

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

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

- 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

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

  is equivalent to the *forward* integration of

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

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

Observation 3

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

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

- 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

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

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

Example: LQR of a pure inertia system: analysis

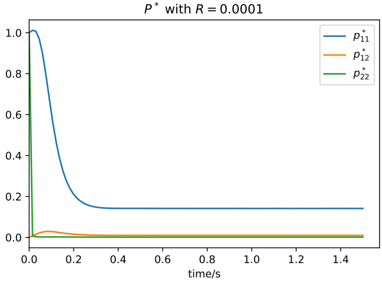


Figure: LQ example: $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $P(t) = P^*(t_f - t)$

- 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

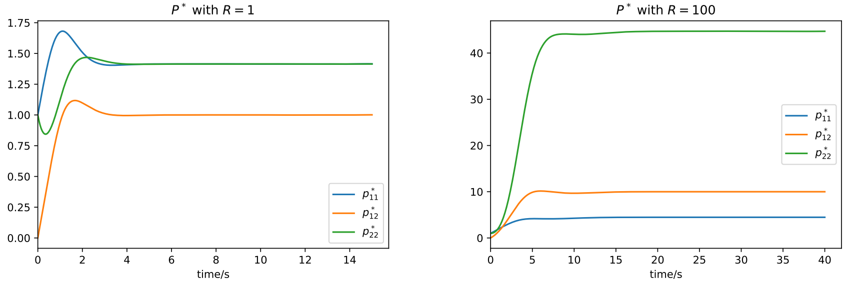
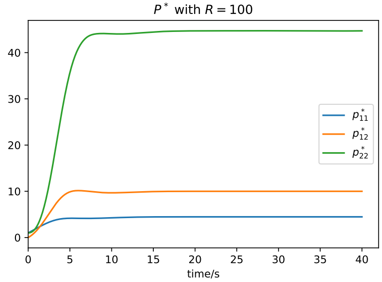


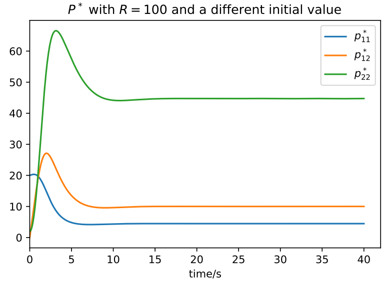
Figure: LQ example with different penalties on control. $P^*(0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

- 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.

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

1. Problem formulation

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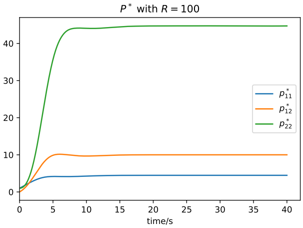
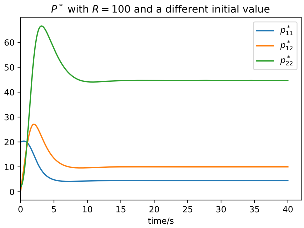
From LQ to stationary LQ





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

Need for controllability/stabilizability

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

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

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

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 $J^0 = \frac{1}{2}x_0^TP(0)x_0$ 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

- 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

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

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

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 $x^TQx = x^TC^TCx$ 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)

From LQ to stationary LQ

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

More formally: Solution of the infinite-horizon LQ

For

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

with $\dot{x}(t) = Ax\left(t\right) + Bu\left(t\right),\ x\left(t_0\right) = x_0$ 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

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

Observations

- the control $u(t) = -R^{-1}B^TPx(t)$ is a *constant* 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

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

- when the ARE solution P_+ is positive definite, $\frac{1}{2}x^TP_+x$ is a Lyapunov function for the closed-loop system

Observations

- Lyapunov Eq. and the ARE:

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

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

- 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:

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

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

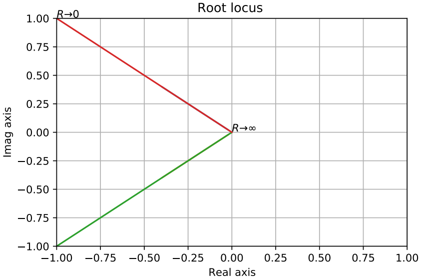


Figure: Eigenvalue $\lambda_{1,2} = -\frac{1}{\sqrt{2}R^{1/4}} \pm \frac{1}{\sqrt{2}R^{1/4}} j$ evolution (root locus)

- $R \uparrow$ (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.

- *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

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

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;
- 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.