

Introduction to Linear Quadratic Optimal Control

MEM 355 Performance Enhancement of Dynamical Systems

Harry G. Kwatny
Department of Mechanical Engineering & Mechanics
Drexel University

Outline

- State Feedback Regulators
 - The LQR Problem
 - Solving the Riccati Equation
- Output Feedback Regulation
 - Coping with Measurement Noise & Disturbances
 - Classical LQG

Regulation: Steer the system from any initial state to the origin.

The Linear Quadratic Regulator Problem (state feedback)

Given : $\dot{x} = Ax + Bu$, Find a state feedback control $u = Kx$
that minimizes the cost function

$$J(x) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T [x^T(\tau)Qx(\tau) + u^T(\tau)Ru(\tau)] d\tau$$

Theorem: If (A, B) is controllable (stabilizable) the optimal controller
for any initial state is $u(t) = Kx(t)$, where:

$$K = -R^{-1}B^T X$$

and $X = X^T \geq 0$ is the unique positive semidefinite solution
of the algebraic Riccati equation:

$$A^T X + XA - XBR^{-1}B^T X + Q = 0$$

Stability of LQR~1

Show that the optimal state feedback control results in a stable closed loop system.

An autonomous linear system $\dot{x} = Ax$ is asymptotically stable if there exists $P > 0, Q > 0$ that satisfy the **Lyapunov equation**

$$PA + A^T P = -Q$$

Furthermore, if $\det A \neq 0$, the system is asymptotically stable if and only if there exist, $P > 0, Q \geq 0$ that satisfy the Lyapunov equation.

Basis of proof: set $V(x) = x^T Px$, $P > 0$ and compute

$$\dot{V} = \frac{\partial V}{\partial x} \dot{x} = x^T PAx + x^T A^T Px = x^T (PA + A^T P)x$$

If $\dot{V} = -x^T Qx$, $Q > 0$ we have stability by the Lyapunov stability Theorem. If $\det A \neq 0$, use LaSalle's Thm for sufficiency, Chetaev Instability Thm for necessity.

Stability of LQR ~ 2

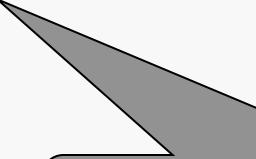
Now consider, $\dot{x} = Ax + Bu, u = Kx \Rightarrow \dot{x} = (A + BK)x$ where

$$K = -R^{-1}B^T X$$

$$XA + A^T X - XBR^{-1}B^T X = -Q$$

$$\Downarrow \quad +XBK - XB(-R^{-1}B^T X) + (BK)^T X - (B(-R^{-1}B^T X))^T X$$

$$X(A + BK) + (A + BK)^T X = -Q - XBR^{-1}B^T X$$



Lyapunov Equation

Generalization of LQR

$$J(x) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \left[x^T(\tau) Q x(\tau) + 2x^T(\tau) S u(t) + u^T(\tau) R u(\tau) \right] d\tau$$

$$R > 0, Q - SR^{-1}S \geq 0$$

Notice that (complete the square)

$$x^T Q x + 2x^T S u + u^T R u = \left(u + R^{-1} S x \right)^T R \left(u + R^{-1} S x \right)^T + x^T \left(Q - SR^{-1} S^T \right) x$$

Define a new control

$$\tilde{u} = u + R^{-1} S^T x, \quad \dot{x} = Ax + Bu \Rightarrow \dot{x} = \left(A - BR^{-1} S^T \right) x + B\tilde{u}$$

$$J(x) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \left[x^T(\tau) \left(Q - SR^{-1} S^T \right) x(\tau) + \tilde{u}^T(\tau) R \tilde{u}(\tau) \right] d\tau$$

$$\Rightarrow \boxed{\begin{aligned} u &= -R^{-1} \left(B^T P + S \right) x \\ &P \left(A - BR^{-1} S^T \right) + \left(A - BR^{-1} S^T \right)^T P - PBR^{-1}B^T P + \left(Q - SR^{-1} S^T \right) \end{aligned}}$$

RE & Associated Hamiltonian Matrix

$$A^T X + X A - X B R^{-1} B^T X + Q = 0 \Rightarrow H = \begin{bmatrix} A & -B R^{-1} B^T \\ -Q & -A^T \end{bmatrix}$$

- if λ is an eigenvalue of H , so is $-\lambda$
- (A, B) controllable, (A, C) observable where $Q = C^T C$
 \Rightarrow there are no eigenvalues on the imaginary axis

$$\exists T = \begin{bmatrix} T_{11} & T_{12} \\ T_{21} & T_{22} \end{bmatrix} \text{ such that } T^{-1} H T = \begin{bmatrix} -\Lambda & 0 \\ 0 & \Lambda \end{bmatrix}$$

Λ is composed of real Jordan blocks, with $\operatorname{Re} \lambda(\Lambda) > 0$

The positive definite solution is $X = T_{21} T_{11}^{-1}$

Computational Tools

LQR Linear-quadratic regulator design for continuous-time systems.

$[K, S, E] = \text{LQR}(A, B, Q, R, N)$ calculates the optimal gain matrix K such that the state-feedback law $u = -Kx$ minimizes the cost function

$$J = \text{Integral} \{x'Qx + u'Ru + 2*x'Nu\} dt$$

subject to the state dynamics $x = Ax + Bu$.

The matrix N is set to zero when omitted. Also returned are the Riccati equation solution S and the closed-loop eigenvalues E :

-1

$$SA + A'S - (SB+N)R^{-1}(B'S+N') + Q = 0, \quad E = \text{EIG}(A-B^*K).$$

Example: Boeing 747 Longitudinal Dynamics

Consider the longitudinal dynamics of a Boeing 747 at 20,000 ft with a speed of 830 fps

$$\begin{bmatrix} \dot{u} \\ \dot{w} \\ \dot{q} \\ \dot{\theta} \\ \dot{h} \end{bmatrix} = \begin{bmatrix} -0.00643 & 0.0263 & 0 & -32.2 & 0 \\ -0.0941 & -0.624 & 820 & 0 & 0 \\ -0.000222 & -0.00153 & -0.668 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & -1 & 0 & 830 & 0 \end{bmatrix} \begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix} + \begin{bmatrix} 0 \\ -32.7 \\ -2.08 \\ 0 \\ 0 \end{bmatrix} \delta_e, y = \theta$$

$\begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix}$	body x velocity
$\begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix}$	body z velocity
$\begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix}$	pitch rate
$\begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix}$	pitch
$\begin{bmatrix} u \\ w \\ q \\ \theta \\ h \end{bmatrix}$	altitude

Boeing 747 Open Loop Modes

```
A=[-0.0064 0.0263 0 -32.2 0;-0.0941 -0.624 761 -196.2 0;  
-0.0002 -0.0015 -4.41 -12.48 0;0 0 1 0 0;0 -1 0 830 0];
```

```
[V,E]=eigs(A)
```

V =

```
0.0032 + 0.0009i 0.0032 - 0.0009i 0.0252 + 0.0000i 0.0673 + 0.0000i 0.0000 + 0.0000i  
0.9985 + 0.0000i 0.9985 + 0.0000i -0.4297 + 0.0000i -0.0104 + 0.0000i 0.0000 + 0.0000i  
-0.0018 + 0.0039i -0.0018 - 0.0039i -0.0000 + 0.0000i -0.0000 + 0.0000i 0.0000 + 0.0000i  
0.0011 - 0.0002i 0.0011 + 0.0002i 0.0001 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i  
-0.0294 + 0.0449i -0.0294 - 0.0449i -0.9026 + 0.0000i -0.9977 + 0.0000i 1.0000 + 0.0000i
```

E =

```
-2.2487 + 2.9839i 0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i  
0.0000 + 0.0000i -2.2487 - 2.9839i 0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i  
0.0000 + 0.0000i 0.0000 + 0.0000i -0.5325 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i  
0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i -0.0105 + 0.0000i 0.0000 + 0.0000i  
0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i
```

Example: Boeing 747 Longitudinal

```
>> A=[-0.0064 0.0263 0 -32.2 0;-0.0941 -0.624 761 -196.2 0;  
     -.0002 -0.0015 -4.41 -12.48 0;0 0 1 0 0;0 -1 0 830 0];  
>> b=[0;-32.7;-2.08;0;0];  
>> Q = [1,0,0,0,0;0,1,0,0,0;0,0,1,0,0;0,0,0,1,0;0,0,0,0,0.000001];  
>> [K,S,E] = lqr(A,b,0.001*Q,1)  
K =  
    0.0253  -0.0178  -2.3696  -4.1886  -0.0000  
E =  
    -5.1565 + 5.4556i  
    -5.1565 - 5.4556i  
    -0.1191 + 0.0932i  
    -0.1191 - 0.0932i  
    -0.0001 + 0.0000i
```

$-5.1565 \pm j5.4556$	$-2.2487 \pm j2.9839$
$-0.1191 \pm j0.0932$	-0.5325
	-0.0105
-0.0001	0
closed loop	open loop

Boeing 747 Cont'd

```
>> [V,Eig]=eig(A-b*K)
```

```
V =
```

```
0.0019 + 0.0007i 0.0019 - 0.0007i -0.0371 - 0.0027i -0.0371 + 0.0027i 0.0010 + 0.0000i  
0.9997 + 0.0000i 0.9997 + 0.0000i 0.0148 - 0.0364i 0.0148 + 0.0364i -0.0002 + 0.0000i  
-0.0052 + 0.0079i -0.0052 - 0.0079i 0.0000 - 0.0000i 0.0000 + 0.0000i 0.0000 + 0.0000i  
0.0012 - 0.0002i 0.0012 + 0.0002i -0.0001 + 0.0001i -0.0001 - 0.0001i -0.0000 + 0.0000i  
-0.0200 + 0.0141i -0.0200 - 0.0141i 0.9985 + 0.0000i 0.9985 + 0.0000i 1.0000 + 0.0000i
```

Linear Quadratic Gaussian (LQG) Problem

Consider the linear system

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

$$y = Cx + v$$

where w, v are independent, zero-mean white noise processes

$$E\{w(t)\} = 0, E\{v(t)\} = 0$$

having covariances

$$E\{w(t)w^T(\tau)\} = W\delta(t-\tau), E\{v(t)v^T(\tau)\} = V\delta(t-\tau), E\{w(t)v^T(\tau)\} = 0$$

Find $u(t)$ that minimizes

$$J = E \left\{ \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T [x^T Q x + u^T R u] dt \right\}, Q = Q^T \geq 0, R = R^T > 0$$

LQG Solution

$$u(t) = K\hat{x}(t), \quad K = -R^{-1}B^T P$$

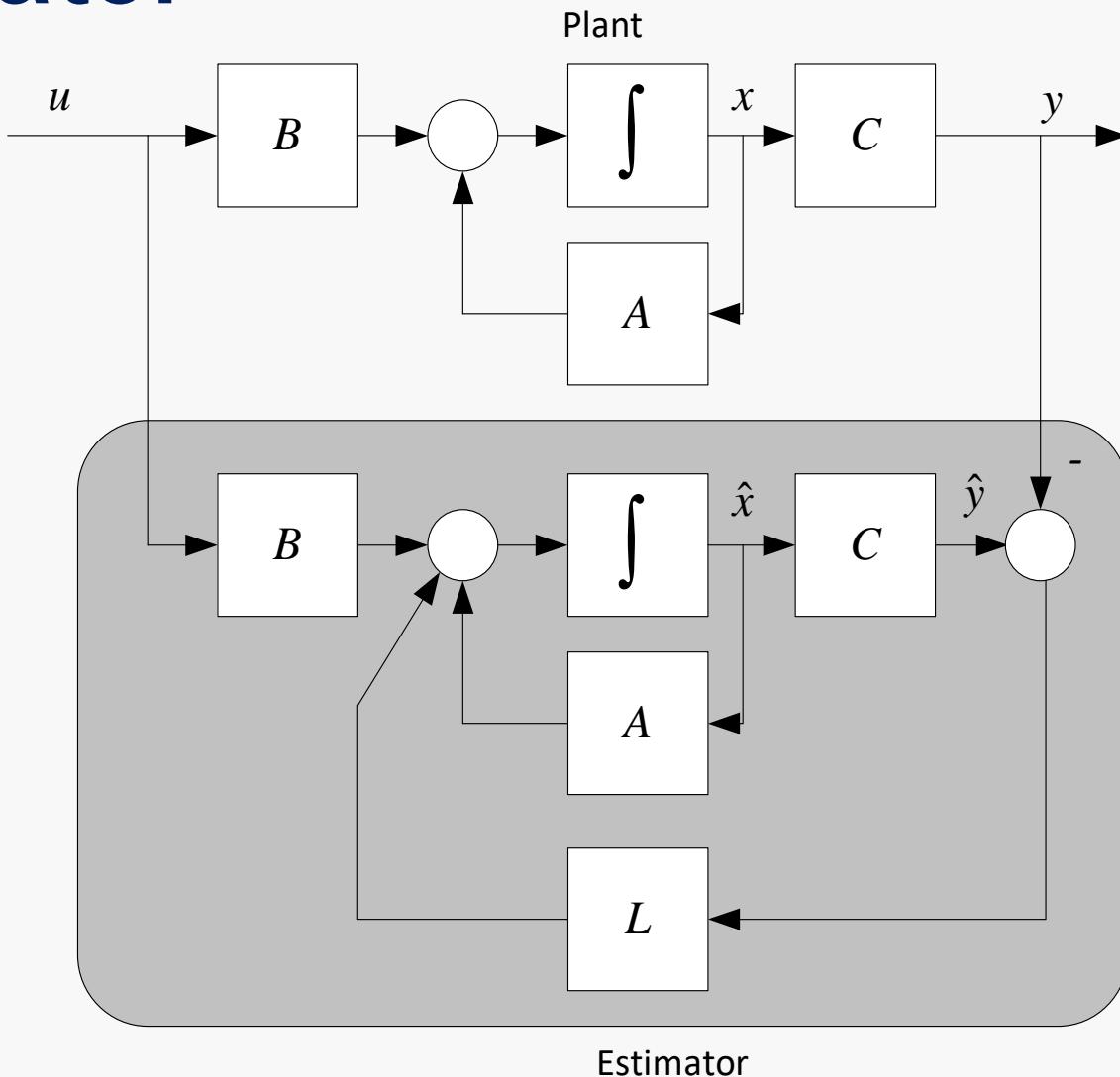
$$PA + A^T P - PBR^{-1}B^T P = -Q, \quad P \geq 0$$

$$\dot{\hat{x}} = A\hat{x} + Bu + L(y - C\hat{x}), \quad L = SC^T V^{-1}$$

$$SA^T + AS - SC^T V^{-1} CS = -W, \quad S \geq 0$$

Full-State Estimator

Normally all states are not Available for measurement. Consequently we design a “state estimator” or “observer” which provides estimates based on the measured plant outputs.



Estimator Dynamics

The estimator design approach is to select L by solving the Riccati Eq

$$SA^T + AS - SC^T V^{-1} CS = -W$$

which compares to

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

with

$$R \Leftrightarrow V, Q \Leftrightarrow W, B \Leftrightarrow C^T$$