

LQR - Problem. is an Optimal Control Problem.

(Linear Quadratic Regulator)

- Optimal Control Problem (Typical form)

minimize Cost functional

$$J = \underbrace{\int_0^{t_f} g(x(t), u(t), t) dt}_{\text{Cost accumulated by running system in interval } [0, t_f]} + \underbrace{h(x(t_f), t_f)}_{\text{Cost incurred at the terminal time.}}$$

Cost accumulated by running system in interval  $[0, t_f]$

Cost incurred at the terminal time.

Subject to system differential equation.

$$\dot{x}(t) = f(x(t), u(t), t)$$

$$x(0) = x_0, \text{ free}$$

$$t_f \text{ ~~free~~ fixed.}$$

- Objective is to choose input  $u(t)$  s.t  $J$  is minimized.

- Soft constraints can be imposed on <sup>Control</sup> System operation using cost functional of type described above

- LQR specifies a cost functions of the form.

$$J = \int_0^{t_f} [x(t)^T Q x(t) + u(t)^T R u(t)] dt + x(t_f)^T N x(t_f)$$

where  $Q, R, N > 0$  are positive definite matrices

- One can tune  $Q, R, N$  to choose controller  $u(t)$  for specific speed of convergence to set points, etc.

One can solve the optimal control problem using Pontryagin's Minimum Principle. Let us see what it is.

Thm: [Pontryagin's Minimum Principle]

For the optimal control  $u^*(t)$  & optimal state trajectory  $x^*(t)$ , there exists non-zero  $p(t)$  s.t. the following equations are satisfied.

$$(i) \frac{dp}{dt} = - \frac{\partial H}{\partial x}^T$$

$$(ii) \dot{x} = \frac{\partial H}{\partial p}$$

$$(iii) p^T(t_f) = \left. \frac{\partial h}{\partial x} \right|_{t=t_f}$$

$$\text{and } u^*(t) = \arg \min_u H(x, p, u, t)$$

$$\begin{aligned} \text{where } H(x, p, u, t) &= \cancel{p^T f(x, u, t) + p^T R u + p^T f(x, u, t)} \\ &= g(x, u, t) + p^T f(x, u, t) \end{aligned}$$

We apply it to LQR problem. (finite horizon)

$$\min J = \int_0^{t_f} \frac{1}{2} (\dot{x}^T Q \dot{x} + u^T R u) dt + \frac{1}{2} x^T(t_f) N x(t_f)$$

$$\text{s.t. } \dot{x} = Ax + Bu$$

$x_0$  specified.

$t_f$  fixed.

Apply necessary conditions:

$$H(x, u, p; t) = \frac{1}{2} (\dot{x}^T Q \dot{x} + u^T R u) + p^T (Ax + Bu)$$

$$(i) \quad \dot{p} = -\frac{\partial H}{\partial p} = -A^T p - Qx$$

$$(ii) \quad \dot{x} = Ax + Bu.$$

$$(iii) \quad p(t_f) = \frac{\partial h}{\partial x(t_f)} = Nx(t_f)$$

To find

$$u^* = \arg \min_u H(x, u, p, t)$$

use

$$\frac{\partial H}{\partial u} = 0 \Rightarrow u^T R + p^T B = 0.$$

$$\Rightarrow u = -R^{-1} B^T p$$

$$\begin{bmatrix} \dot{x} \\ \dot{p} \end{bmatrix} = \begin{bmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} x \\ p \end{bmatrix}, \quad \begin{matrix} x(0) = x_0 \\ p(t_f) = Nx(t_f) \end{matrix} \quad \left. \vphantom{\begin{bmatrix} \dot{x} \\ \dot{p} \end{bmatrix}} \right\} \begin{matrix} \text{Althang } x(t_f) \\ \text{is free} \\ \text{Two point} \\ \text{Boundary} \\ \text{Value problem} \end{matrix}$$

from (iii) we guess that nature of  
 $p(t) = K(t)x(t)$  where  $K(t)$  is matrix  
function of time.

and at  $t = t_f$ ,  $K(t_f) = N$ .

Thus  $u(t) = -R^{-1}B^T K(t)x(t)$

also

$$\begin{aligned}\dot{p}(t) &= \dot{K}(t)x(t) + K(t)\dot{x}(t) \\ &= \dot{K}x + K(Ax + Bu)\end{aligned}$$

$$-Qx - A^T p = \dot{K}x + KAx + KBu$$

$$-Qx - A^T Kx = \dot{K}x + KAx + KBR^{-1}B^T Kx$$

$$\Rightarrow \dot{K} + KA + A^T K + Q - KBR^{-1}B^T K = 0$$

since  $x(t) \neq 0$ .

$$\dot{K} + KA + A^T K + Q - KBR^{-1}B^T K = 0 \quad \text{--- (1)}$$

(Riccati Differential Equation)

with  $K(t_f) = N$ .

Thus to solve finite horizon LQR problem  
one gets a time varying state-feedback  
gain matrix <sup>using</sup>  $K(t)$  as follows

Solve (1) for  $K(t)$

use  $u^*(t) = -R^{-1}B^TK(t)x(t)$

as optimal controller.

- Infinite Horizon case.

~~$K(t)$~~  ~~As we solve~~ ~~very~~ ~~is~~  
As we ~~start~~ move backwards from

~~$t_f$~~ ,  $K(t_0)$  quickly converges to

equilibrium [See Donald Kirk or any other  
Optimal control book]

Roughly arguing <sup>(because of above)</sup> in infinite horizon case

we need solution to Algebraic Riccati  
equation which is equilibrium solution to  
Differential Riccati eqn. (1).

$$A^TK + KA + Q - KBR^{-1}B^TK = 0$$

ex:  $\dot{x} = x + u$ . find  $u$  to:

$$\min J = \int_0^1 \frac{1}{2}(x^2 + u^2) dt + \frac{1}{2}x(1)^2$$

$$Q = 1, \quad R = 1, \quad N = 1$$

$$\dot{K} = +2K - K^2 + 1$$

$$K(1) = 1$$

Problem

$$\text{Solve it } \int \frac{dK}{2K - K^2 + 1} = \int dt = t + c.$$

find  $c$  using  $K(1) = 1$ .

Controllability Gramian  $J = \int_0^{\infty} u^T u dt$

$$\text{ARE: } A^T K + K A - K B R^{-1} B^T K = 0.$$

put  $P = K^{-1}$  and multiply on both sides  
(we seek symmetric  $K$ )

$$P A^T + A P = B B^T \quad (\text{Lyapunov eqn with } Q = B B^T)$$

Solution to this eq. is Controllability Gramian.

$$P = \int_0^{\infty} e^{A^T t} B B^T e^{A t} dt$$