

1 Dynamic programming

- Key concept : solve a (usually large & difficult) problem via solving a collection of sub problems.

- Bellman's principle of optimality

"From any point on an optimal trajectory, the remaining trajectory is optimal for the corresponding problem initiated at that point".

- In math : Bellman Equation

$$J(s) = \min (J(s') + \text{Reward}(s' \rightarrow s))$$

2 Discrete time LQ (Regulation)

- System :
$$\begin{aligned} x(k+1) &= A x(k) + B u(k) \\ y(k) &= C x(k) \end{aligned}$$

- performance index (Cost function):

$$J = \frac{1}{2} x^T(N) S x(N) + \frac{1}{2} \sum_{k=k_0}^{N-1} x^T(k) Q x(k) + u^T(k) R u(k)$$

S, R, Q (PD)

$$J^0 = \min_{u(k_0) \dots u(N-1)} J$$

- With dynamic programming.

$$J_k^0(x(k)) = \min_{u(k)} \left\{ \frac{1}{2} x^T(k) Q x(k) + \frac{1}{2} u^T(k) R(k) u(k) + J_{k+1}^0(x(k+1)) \right\}$$

boundary condition : $J_N^0(x(N)) = \frac{1}{2} x^T(N) S x(N)$

- Solution (How to introduce this?)

$$u(k) = - \underbrace{\{ [R + B^T P(k+1) B]^{-1} B^T P(k+1) A \}}_{K(k)} x(k)$$

$J_{k_0}^0(x(k_0)) = \frac{1}{2} x_0^T P(0) x_0$; $J_k^0(x(k)) = \frac{1}{2} x^T(k) P(k) x(k)$

- Riccati Equation :

$$P(k) = A^T P(k+1) A - A^T P(k+1) B [R + B^T P(k+1) B]^{-1} B^T P(k+1) A + Q$$

$$S = P(N)$$

- Implementation
 1. $P(k)$ is usually generated offline
 2. $u(k)$ is computed online.

For infinite LQ: (First assume P_k will converge).

Algebraic Riccati equation

$$P_s = A^T P_s A + Q - A^T P_s B [R + B^T P_s B]^{-1} B^T P_s A$$

$$u^o(k) = - \underbrace{[R + B^T P_s B]^{-1} B^T P_s A}_{K_s} x(k)$$

Stability proof:

1 If (A, B) are stabilizable and $S=0$, riccati equation converge to a bounded solution $P_s > 0$

(a) $S=0$, $J^o(N) = \min_u \left\{ \frac{1}{2} \sum_{i=1}^N (x^T(i) Q x(i) + u^T(i) R u(i)) \right\}$ is non-decreasing function on horizon N .

(b) (A, B) is stabilizable, there exist a control make $x_c(k) \rightarrow 0$ geometrically, $\Rightarrow \sum_{k=0}^{\infty} \|x_c(k)\|^2 < \infty$
 $\Rightarrow J^o(\infty)$ is bounded

(c) $J^o(N)$ is monotone and bounded, $J^o(N) = J^o(\infty)$
~~Since~~ $J^o(\infty) < \infty$,

$$J^o(\infty) = \frac{1}{2} x(0)^T P x(0) \Rightarrow P \text{ is bounded.}$$

$$X(k+1) = (A - BK_S) X(k) = A_{cl} X(k)$$

Observability $\Rightarrow P_S > 0$ with $u^0(k) = -K_S X(k)$

$$Q = C^T C$$

$$J = X_0^T P_S X_0 = \sum_{k=0}^{\infty} \left\{ \underbrace{X^T(k) Q X(k)}_{Q = C^T C} + \underbrace{u^T(k) R u(k)}_{u = -K_S X(k)} \right\}$$

$$= \sum_{k=0}^{\infty} \left\{ X^T(k) \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix}^T \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix} X(k) \right\}$$

$$= \sum_{k=0}^{\infty} \left\{ X_0^T (A_{cl}^k)^T \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix}^T \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix} X(k) \right\}$$

$$= X_0^T \underbrace{W_{cl}}_{\text{observability gramian for } \begin{matrix} X(k+1) = A_{cl} X(k) \\ \tilde{y}(k) = \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix} X(k) \end{matrix}} X_0$$

$$W_{cl} = \sum_{k=0}^{\infty} (A_{cl}^k)^T \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix}^T \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix} A_{cl}^k$$

$$X(k+1) = (A - BK_S) X(k) = A_{cl} X(k) = AX - Bu.$$

$$\tilde{y}(k) = \begin{bmatrix} C \\ R^{\frac{1}{2}} K_S \end{bmatrix} X(k)$$

$$u(k) = \begin{bmatrix} 0 & -R^{\frac{1}{2}} \end{bmatrix} \tilde{y}(k)$$