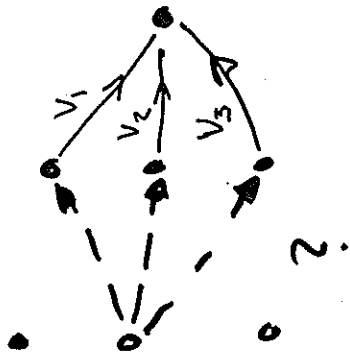


Dagens idé:



Dynamisk programmering

LOG

- Formulering
- Förarbeten
- Fullst. tillståndsinfo. tidsdiagram
- Ofullst. - " -
- Separation
- Tidskont. fallet

Problem

$$\begin{cases} x(t+1) = \phi x(t) + \Gamma u(t) + v(t) \\ y(t) = \theta x(t) + e(t) \end{cases}$$

$$\begin{matrix} R_1 \\ R_2 < 0 \\ R_2 \end{matrix}$$

m, R_0 givet

Minimera

$$J = EL = E \left\{ x^T(N) Q_0 x(N) + \sum_{t=t_0}^{N-1} (x^T(t) Q_1 x(t) + \bar{u}^T(t) Q_2 \bar{u}(t)) \right\} \geq 0$$

Om Q_{12}

$$2 x^T(t) Q_{12} u(t)$$

$$\bar{u} = u + \Gamma^T x$$

$$\Pi = Q_{12} Q_2^{-1}$$

$$\tilde{x}(t+1) = \tilde{\phi} x(t) + \tilde{\Gamma} u(t) + v(t)$$

$$y(t) = \theta x(t) + e(t)$$

$$\tilde{\phi} = \phi - \Gamma \Pi^T$$

$$\tilde{Q}_1 = Q_1 - Q_{12} Q_2^{-1} Q_{12}^T$$

Fria parametrar

$$\begin{matrix} Q_1 & Q_{12} & Q_2 \\ & & \geq 0 \end{matrix}$$

Tidskont. förlustfunktion

$$J = E \left\{ \int_0^{N_h} [x^T Q_{1c} x + 2 x^T Q_{2c} u + \bar{u}^T Q_{2c} \bar{u}] dt + x^T(N_h) Q_{0c} x(N_h) \right\}$$

Integrera över ett sampling-intervall

Styrlagar

- $y(t) = x(t)$ Fallst tillst. info $u(x, t)$
- Öppet. tillst. info. $u(y_{t-1}), u(y_t)$

Några lemma

Lemma 3.1

$$\min_{u(x,y)} EL(x,y,u) = E \min_m L(x,y,m)$$

Lemma 3.2

$$\min_{u(y)} EL(x,y,u) = E \left\{ \min_y E \{ L(x,y,u) | y \} \right\}$$

$$\min_{u(y)} EL \geq \min_{u(x,y)} EL$$

Lemma 3.3

$$E x^T S x = m^T S m + \text{tr}(SR)$$

$$x \in N(m, R)$$

$$E x^T S x = E (x-m)^T S (x-m) + m^T S m$$

$$(x-m)^T S (x-m) = \text{tr} \left(\begin{matrix} & \\ & \end{matrix} \right)^T S \left(\begin{matrix} & \\ & \end{matrix} \right) = \text{tr} S (x-m)(x-m)^T$$

$$E (x-m)^T S (x-m) = \text{tr}(SR)$$

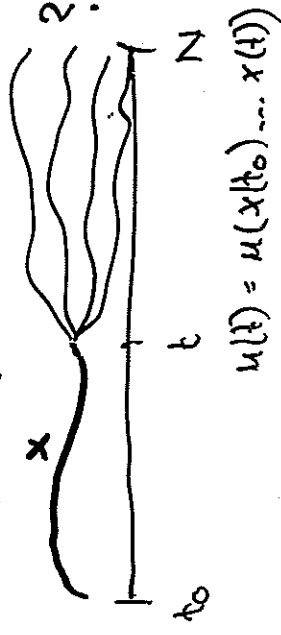
5

Fullständig tillst. info

Dynamisk programmering (Bellman)

Optimalitetsprincipen

Bygger på Hamilton-Jacobi formul.



Hur påverkar $u(t)$
framtida $x(t)$?

Förlustfunktionen

$$J = \min_t E \left\{ \sum_{t=0}^{t-1} x^T Q_1 x + u^T Q_2 u \right\}$$

— översikt av framtiden

$$\rightarrow \min_t E \left\{ x^T(n) Q_0 x(n) + \sum_t^{N-1} x^T Q_1 x + u^T Q_2 u \right\}$$

J_f

Lemma 3.1

$$J_f = E \min_{u(t), \dots, u(N-1)} \left\{ x_0^T Q_0 x_0 + \sum_t^{N-1} x^T Q_1 x + u^T Q_2 u \right\}$$

$V(x, t)$

Bellman eqn

$$\begin{aligned} V(x(t), t) = & \min_{u(t)} E \left[x^T(t) Q_1 x(t) + u^T(t) Q_2 u(t) \right] \\ & + \min_{u(t+1)} E \left[x^T(t+1) Q_1 x(t+1) + u^T(t+1) Q_2 u(t+1) \right] \\ & + \min_{u(t+2)} E \left[\dots x(t+2) x(t+1) x(t) \right] \end{aligned}$$

$\vdots$

$$\begin{aligned} V(x(t), t) = & \min_{u(t)} \left[x^T(t) Q_1 x(t) + u^T(t) Q_2 u(t) \right. \\ & \left. + E \left\{ V(x(t+1), t+1) | x(t) \right\} \right] (*) \end{aligned}$$

Initialvärde

$$V(x, N) = \min_{u(N)} E \left[x^T(N) Q_0 x(N) | x \right] = x^T Q_0 x$$

Kan visa att lösningen till (*)
kan skrivas på formen

$$V(x, t) = x^T S(t) x + s(t) > 0$$

Använd induktion

Resultat

$$u(t) = -L(t)x(t)$$

$$S(t) = \Phi^T S(t+1) \Phi + Q_1 - L^T (Q_2 + \Gamma^T S(t+1) \Gamma)^{-1} L$$

$$S(N) = Q_0$$

$$\min E \{ \quad \} = E [V(x, t_0) | x]$$

$$= E [x^T(t_0) S(t_0) x(t_0) + s(t_0)]$$

$$= m^T S(t_0) m^T + \text{tr} S(t_0) R_0 \quad \text{Initialv.}$$

$$+ \sum_{t=t_0}^{N-1} \text{tr} R_1 S(t+1) \quad \text{Bms } t_0, \dots, N-1$$

OFULLST. TILLSTÅNDS INFO.

$$u(t) = f(y_{t-1})$$

$$E \left[\sum_{t_0}^{t-1} x^T Q_1 x + m^T Q_2 m \right]$$

$$+ E \left[x^T(N) Q_0 x(N) + \sum_t^{N-1} x^T Q_1 x + m^T Q_2 m \right]$$

$$\min_{u(t)} E \left[x^T(N) Q_0 x(N) + \sum_{st}^{N-1} x^T(s) Q_1 x(s) + m^T(s) Q_2 m(s) \mid y_{t-1} \right]$$

Samma som tidigare

$$V(y_{t-1}, t) = \min_u E [x^T Q_1 x + m^T Q_2 m$$

$$+ V(y_t, t+1) \mid y_{t-1}]$$

Problem med y_t
än för

$$W(\hat{x}(t), t) = V(y_{t-1}, t)$$

$$W(\hat{x}(t), t) = \min_u E[x^T Q_1 x + u^T Q_2 u$$

$$+ W(\hat{x}(t+1), t+1) | \hat{x}(t)]$$

OBS $\hat{x}(t)$ vet vi har vi kan få

Blir samma räkningar som tidigare men lite extra färdigheter

$$\begin{array}{c} m^T S(t_0) m + \text{tr } R_0 S(t_0) + \sum_{k=0}^{N-1} \text{tr } R_1 S(t+k) \\ \xleftarrow{N-1} \xrightarrow{1-C} \xrightarrow{t_0} \xrightarrow{t_0} \xrightarrow{v(t)} \end{array} \quad \begin{array}{c} + \sum_{k=0}^{N-1} \text{tr } P(k) L^T(k) \Gamma^T S(t+k) \phi \\ \xleftarrow{t_0} \xrightarrow{e(t)} \end{array}$$

$$S = S(Q_0, Q_1, Q_2)$$

$$P = P(R_0, R_1, R_2)$$

$$u(t) = -L(t) \hat{x}(t|t-1)$$

$\hat{x}(t|t)$?

SEPARATION

EGENSKAPER

$$x(t+1) = (\Phi - \Gamma L)x(t) + \Gamma L \tilde{x}(t) + v$$

$$\tilde{x}(t+1) = (\Phi - K\Theta) \tilde{x}(t) + v(t) - K e(t)$$

$$\begin{bmatrix} x(t+1) \\ \tilde{x}(t+1) \end{bmatrix} = \begin{bmatrix} \Phi - \Gamma L & \Gamma L \\ 0 & \Phi - K\Theta \end{bmatrix} \begin{bmatrix} x(t) \\ \tilde{x}(t) \end{bmatrix} + \begin{bmatrix} v \\ v(t) - K e(t) \end{bmatrix}$$

Asymptotisk stabilitet:

$$\Phi - \Gamma L$$

$$\Phi - K\Theta$$

asympt. stab. matriser

Tidskontinuerliga fall

12

$$\begin{cases} dx = Ax dt + B u dt + dv \\ dy = Cx dt + de \end{cases}$$

Förlustfunktion

$$E \left\{ x^T(t_1) Q_0 x(t_1) + \int_{t_0}^{t_1} (x^T Q_1 x + u^T Q_2 u) dt \right\}$$

1. Deterministiska fall

13

$$dv = de = R_1 = R_2 = 0$$

$$J = x^T(t_0) S(t_0) x(t_0)$$

$$\mu = -Lx = -Q_2^{-1} B^T S x$$

$$EJ = m^T S(t_0) m + \text{tr } S(t_0) R_0$$

2. Fullständig tillståndsinformation

$$\mu = -Lx$$

$$EJ = m^T S(t_0) m + \text{tr } S(t_0) R_0 + \int_{t_0}^{t_1} \text{tr}(R_1 S) dt$$

3. Ofullst. tillståndsinformation

$$\mu = -L\hat{x}$$

$$EJ = m^T S(t_0) m + \text{tr } S(t_0) R_2 + \int_{t_0}^{t_1} \text{tr}(R_1 S) dt + \int_{t_0}^{t_1} \text{tr}(L^T Q_2 L P) dt$$

Analog

$$-\frac{ds}{dt} = A^T s + SA + Q, -SBQ_2^T B^T s$$

$$S(t_1) = Q_0$$

huv lösning ≥ 0 ; $[t_0, t_1]$ da

$$J = x^T(t_1) Q_0 x(t_1) + \int_{t_0}^{t_1} [x^T Q_1 x + u^T Q_2 u] dt$$

$$= x^T(t_0) S(t_0) x(t_0)$$

$$+ \int (u + Q_2^{-1} B^T S x)^T Q_2 (u + Q_2^{-1} B^T S x) dt$$

$$+ \int \text{tr } R_1 S dt + \int dv^T S x + \int x^T S dv$$

Bevis

$$x^T(t_1) Q_0 x(t_1) = x^T(t_1) S(t_1) x(t_1)$$

$$= x^T(t_0) S(t_0) x(t_0) + \int_{t_0}^{t_1} d(x^T S x)$$

$$d(x^T S x) = dx^T S x + x^T S dx + x^T \frac{ds}{dt} x dt + \text{tr } S R_1 dt$$

Sliding Systemet

$$\begin{bmatrix} \dot{d}x \\ \dot{d}\tilde{x} \end{bmatrix} = \begin{bmatrix} A-BL & BL \\ 0 & A-KC \end{bmatrix} \begin{bmatrix} x \\ \tilde{x} \end{bmatrix} dt + \begin{bmatrix} dv \\ dv-Kde \end{bmatrix}$$

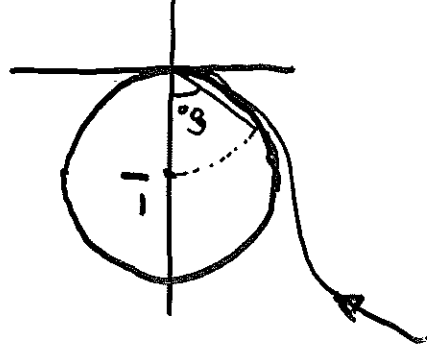
Kräv 1) Stabiliserbarhet $[A, B]$

2) Detekterbarhet $[A, D]$ $DD^T = Q_1$

$\Rightarrow S(\infty)$ existerar entydigt

1) $A-BL$ stabil

2) $0.5 \leq A_m < \infty$ Fallst. x tillg
 $\varphi_m \geq 60^\circ$



3) Med Kalman filter

"None"

4) Cheap control

Guaranteed Margins for LQG Regulators

JOHN C. DOYLE

Abstract—There are none.

INTRODUCTION

Considerable attention has been given lately to the issue of robustness of linear-quadratic (LQ) regulators. The recent work by Safonov and Athans [1] has extended to the multivariable case the now well-known guarantee of 60° phase and 6 dB gain margin for such controllers. However, for even the single-input, single-output case there has remained the question of whether there exist any guaranteed margins for the full LQG (Kalman filter in the loop) regulator. By counterexample, this note answers that question; there are none.

A standard two-state single-input single-output LQG control problem is posed for which the resulting closed-loop regulator has arbitrarily small gain margin.

T-AC Aug -78

SISO fallet

$$y(k) = \frac{B}{A} u$$

$$\Sigma(y^2 + S u^2)$$

Slutna supplement

$$H_c = C(zI - (\Phi - \underline{Q})') \Gamma = \frac{B}{P}$$

där

$$S A(z^{-1}) A(z) + B(z^{-1}) B(z) = r P(z^{-1}) P(z)$$

$$r = \Gamma^T S \Gamma + g$$