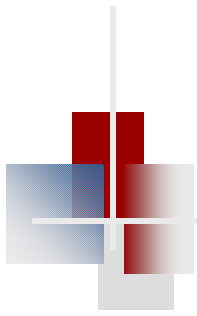


Optimal Control A Game Theoretic Approach

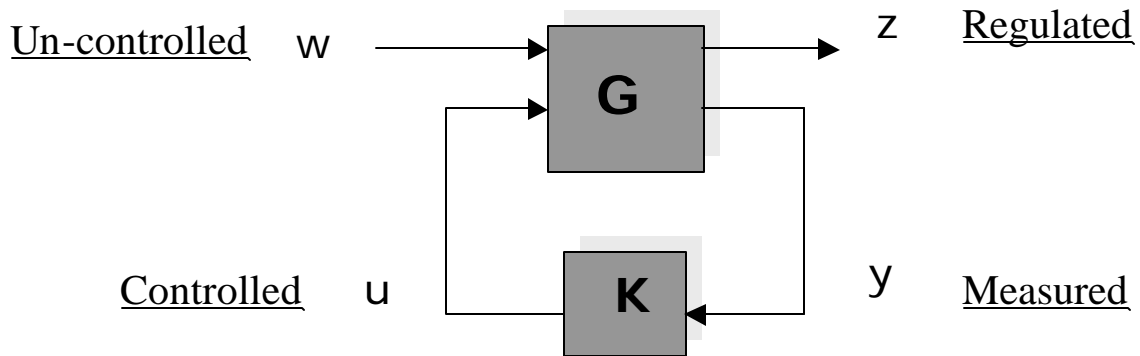
GE 493 – Game Theory

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Robust Control



Robustness

- Stability (stable under perturbations from nominal model)
- Performance (perform satisfactorily under perturbations).

Aim

- Disturbance Rejection/attenuation
- Design u so as to keep z “small”

Game Theoretic Approach

Problem is viewed as a Game between Man (Controller) and Nature (unknown disturbances).



Controller Design Problem

State Equations

$$\dot{x} = A(t)x(t) + B(t)u(t) + D(t)w(t), \quad x(0) = 0$$

$$u(t) = \mathbf{m}(t, x_{[0,t_f]}), \quad t \in [0, t_f]$$

Player 1: Controller

$\mathbf{m} \in M_{cl}$ all closed loop mappings

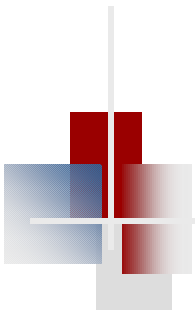
Player 2: Disturbance

$$w \in H_w$$

Cost Function

$$L(u, w) = |x(t_f)|_{Q_f}^2 + \int_0^{t_f} \left\{ |x(t)|_{Q_f}^2 + |u(t)|^2 \right\} dt$$

Linear Quadratic Dynamic Game



Saddle Point

Design a controller such that the gain from w to z is minimized

Minimize:

$$\inf_{\mathbf{m} \in M} \langle \langle T_{\mathbf{m}} \rangle \rangle =: \mathbf{g}^*$$

where

$$\langle \langle T_{\mathbf{m}} \rangle \rangle := \sup_{w \in H_w} \frac{\|T_{\mathbf{m}}(w)\|_z}{\|w\|_w}$$

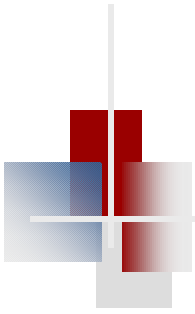
$$L_{\mathbf{g}}(\mathbf{m}, w) = \|T_{\mathbf{m}}(w)\|_z^2 - \mathbf{g}^2 \|w\|_w^2 \equiv \tilde{L}(\mathbf{m}, w) - \mathbf{g}^2 \|w\|_w^2$$

Find smallest value of $\gamma > 0$ such that upper value of the game (with objective function L_{γ}) is bounded

Saddle Point Condition

$$L_{\mathbf{g}}(\mathbf{m}^*, w) \leq L_{\mathbf{g}}(\underbrace{\mathbf{m}^*, w^*}_{\text{SaddlePoint}}) \leq L_{\mathbf{g}}(\mathbf{m}, w^*), \quad \forall \mathbf{m} \in \mathbf{M}, w \in H_w$$

$$\overbrace{\inf_{i \in M} \sup_{w \in H_w} \|T_{\mathbf{m}}(w)\|_z / \|w\|_w}^{\text{Upper Value}} \geq \underbrace{\sup_{w \in H_w} \inf_{i \in M} \|T_{\mathbf{m}}(w)\|_z / \|w\|_w}_{\text{Lower Value}}$$



Saddle Point Solution

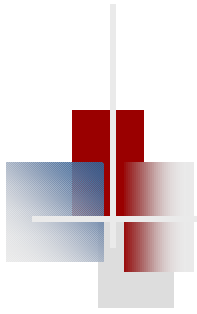
- Introduce a matrix differential equation (Riccati) associated with the game.
- Game admits a unique strongly time consistent saddle-point policy \mathbf{m}^* iff the M.D.E does not have a conjugate point in $[0, t_f]$.
- If above condition is not satisfied

β

No saddle point

β

disturbance will drive L_γ arbitrarily large !!



Measurement Schemes

- Perfect State Measurement.

$$u(t) = \mathbf{m}(t; x(s), s \leq t), \quad t \geq 0$$

- Delayed State Measurement.

$$u(t) = \mathbf{m}(t; x_{0,t-q}], \quad t \geq q$$

- Sampled State Measurement.

$$u(t) = \mathbf{m}(t; x(t_k), \dots, x(t_1), x_0), \quad t_k \leq t \leq t_{k+1}$$



Robustness to Plant Perturbations

System Equations

$$\dot{x} = (A + E)x + Bu, \quad x(0) = x_0 \in \mathfrak{R}^n$$

Plant perturbation

Measurement

$$y = Cx$$

Control

$$u = Ky$$

$$E = \sum_{i=1}^n G_i L_i H_i \quad \text{tr}(L_i^T L_i) \leq \mathbf{d}_i^2$$

Want to stabilize the system for all possible E.

Cost Function

$$J = \int_0^{\infty} (x^T Q x + u^T R u) dt$$

Find K^* and E^* such that :

$$J(K^*, E^*) = \max_E J(K^*, E) = \inf_K \max_E J(K, E)$$

Assumption:

The optimizing solution E^* lies on the boundary



Assumptions & Solution

- Instead of the hard-constrained problem (robust control), an alternate soft-constrained (dynamic game) problem is solved.
- The assumption is that the solution to the hard problem lies on the boundary.

$$\dot{x} = Ax + Bu + \sum_{i=1}^n G_i w_i \quad z_i = H_i x$$

$$J_g \equiv J - \sum_{i=1}^n g_i \text{tr}(L_i^T L_i) \quad w_i = L_i z_i$$

No constraints on L_i

- Solution to the 2nd problem is the the solution to the original problem (under the above assumptions)



Conclusion and Discussion

- Two different robustness issues were dealt with using a Game Theoretic approach.
- Both problems were analyzed in a Quadratic dynamic game framework.
- Different measurement schemes made the analysis more practical (real-life).
- The assumption that the solution exists on the boundary was not justified (no practical examples were provided).