

MTNS 2004

Robust Control: from linear to nonlinear

Performance and robustness analysis and
design for LTI systems: a quick overview
from a robust control point of view

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Outline

- Considered problem
- H_∞
- μ analysis
- Conclusion

Considered problem

Considered problem

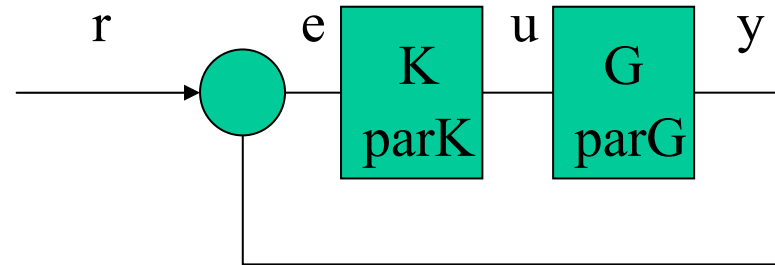
- What is a good method in automatic control ?
- For:
 - Identification
 - Synthesis
 - Analysis
- Looking for methods who are:
 - General
 - Easy to use
 - CAD (computer-aided design)

Considered problem

- **A robustness definition**
 - Let consider a property **P** (stability, raising time, overshoot ...)
 - And a set of models **F**
 - A controller **K** is **P**-robust, if the connexion **K*F** ensures the property **P** for all the models in the set **F**.
- Note that :
 - *Robustness is defined for a given property **P** and a given family **F**.*
 - *Robustness was proposed long time again before 1990!... The major advance was to formalize and explicitly integrate this notion in controller design methods.*

Considered problem

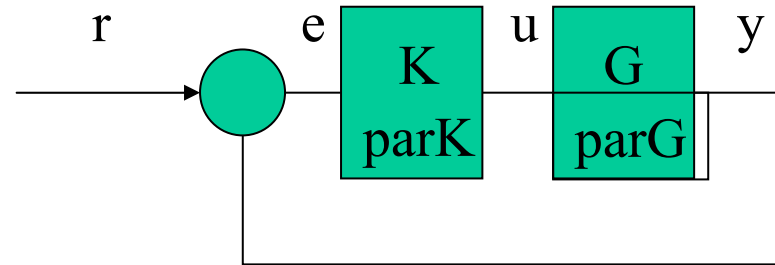
- Some optimisation problem for control



- LQG $\Rightarrow \min_K \int e(t)^2 dt + \int u(t)^2 dt$
Spec. formulation: partial; Resolution: easy
- Structured LQG $\Rightarrow \min_{parK} \int e(t)^2 dt + \int u(t)^2 dt$
Nice problem ... but very difficult

Considered problem

- Some optimisation problem for control



- H_∞ optimisation $\Rightarrow \min_{parK} \text{template} \left(\frac{e(j\omega)}{r(j\omega)} \right)$

Spec. formulation: partial but classical (shaping);
Resolution: easy

- Structured robust Control $\Rightarrow \min_{parK} \max_{parG} J(K, G)$

Spec. formulation: ideal ... just a dream

- Many others (control, analysis, identification...)

Considered problem

- General problems are often **non polynomial problem (NP)**



No well suited for engineering approach

- Even some simple and classical problems. For example, the multivariable gain margin

$$\min_{\text{par}G / \text{BoucleStable}} \|\text{par}G\|$$

- The smallest vector of parameters that destabilize the closed loop.
- NP hard problem

Considered problem

What can we do with the
NP engineer questions ?



General
criterion

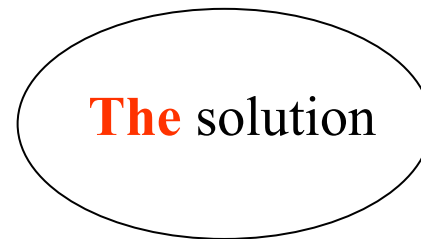
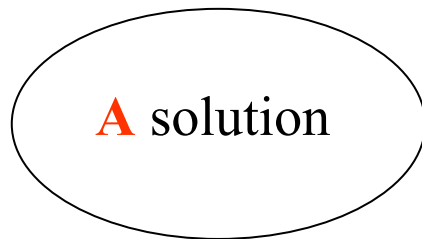
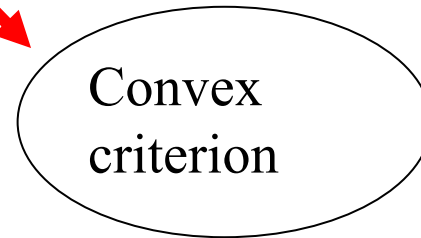
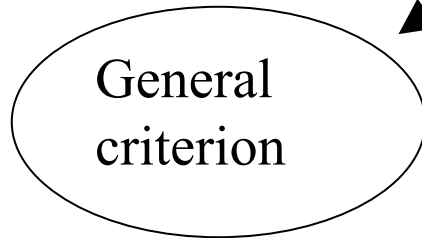


A solution



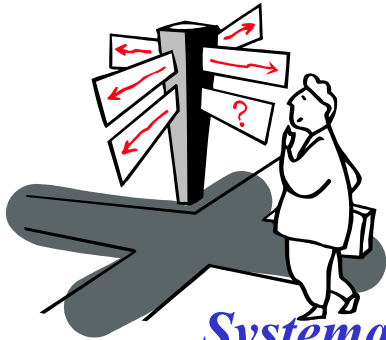
Considered problem

What can we do with the
NP engineer questions ?



Considered problem

What can we do with the
NP engineer questions ?



Systematic translation

General
criterion

Convex
criterion

General optimisation
+ a priori information
Xstart, bounds, variable
transformation...

A solution

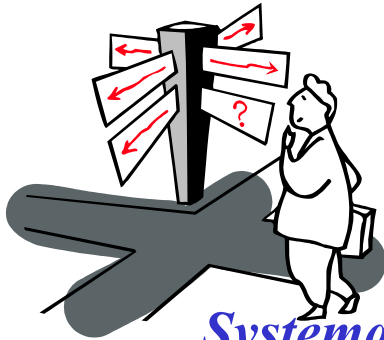
The solution

Local minima

These approaches are not mutually exclusive

Considered problem

What can we do with the
NP engineer questions ?



Systematic translation

Rewrite the specs

General
criterion

Convex
criterion

Immersion
Multiplier
Relaxation

General optimisation
+ a priori information
Xstart, bounds, variable
transformation...

Convex optimisation

A solution

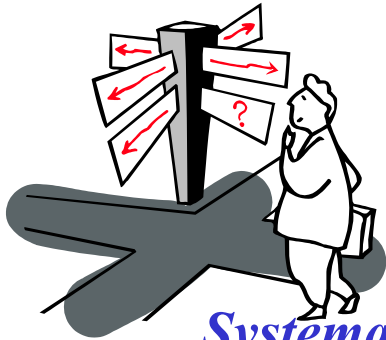
The solution

Local minima

Global minimum

These approaches are not mutually exclusive

Considered problem



What can we do with
NP engineer ques

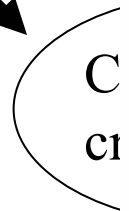
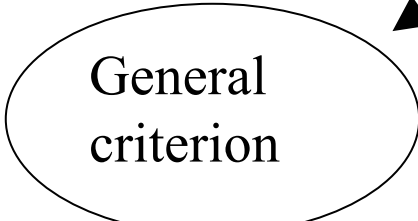
Enlarge the field of robust control methods to nonlinear systems

while preserving:

- global properties for stability
- global performance specs
- direct consideration of robustness specs

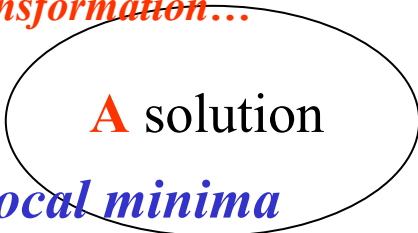
Systematic translation

Rew



General optimisation
+ a priori information
Xstart, bounds, variable transformation...

Convex optimisation

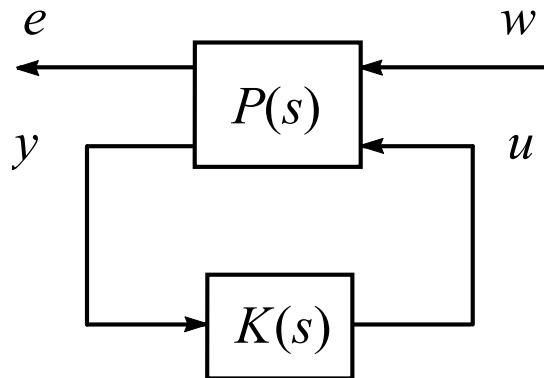


Those approach are not mutually exclusive

H_∞

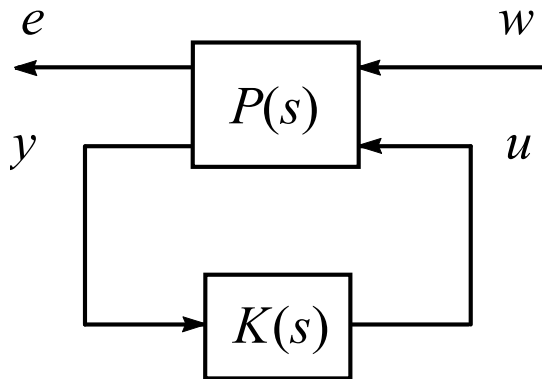
H_∞ – Standard problem

H_∞ – Standard problem



- w : input signals (perturbations, noises, ...)
- u : control signals
- e : output signals that allows characterization of a ‘good’ behaviour (typically look like an ‘error’ signal)
- y : measurement signals

H_∞ – Standard problem

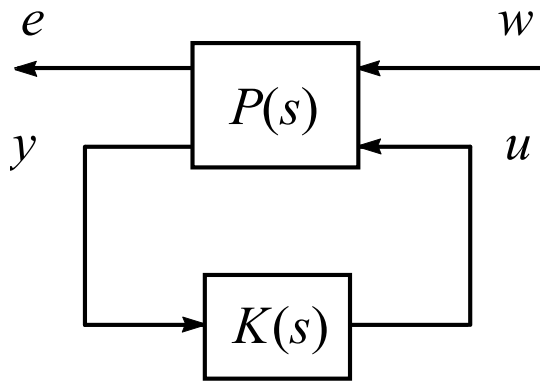


$$\| \mathbf{F}_l(P(s), K(s)) \|_\infty < \gamma$$

- **standard problem**
 - $P(s)$ and γ
 - Find $K(s)$ such as
 - closed loop is stable
 - and $\|Te/w\|_\infty \leq \gamma$

- Controllers that allow to obtain the smallest value γ^* are optimal.

H_∞ – Standard problem



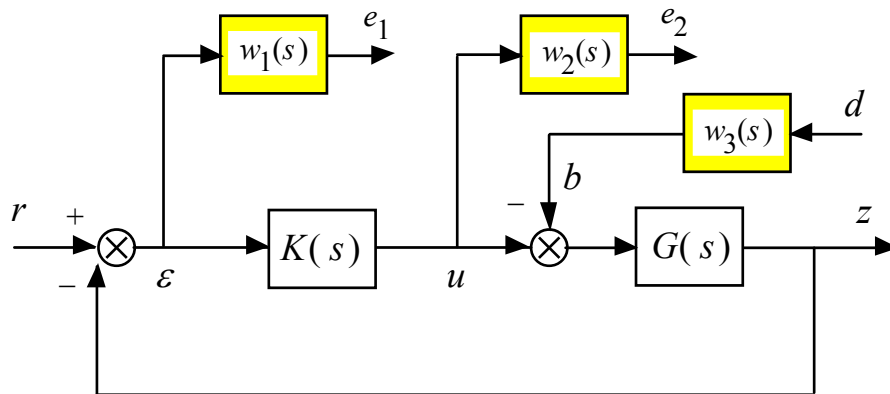
- Solutions:

- Hankel operator (decomposition, projection,...)
- Glover-Doyle solution at the end of the 80's using Riccati equations
- Kwakernak polynomial approach give nice insight on the behaviour of central solution and rank deficiency of the optimal controller
- Gahinet, Apkarian method (1994) using LMI

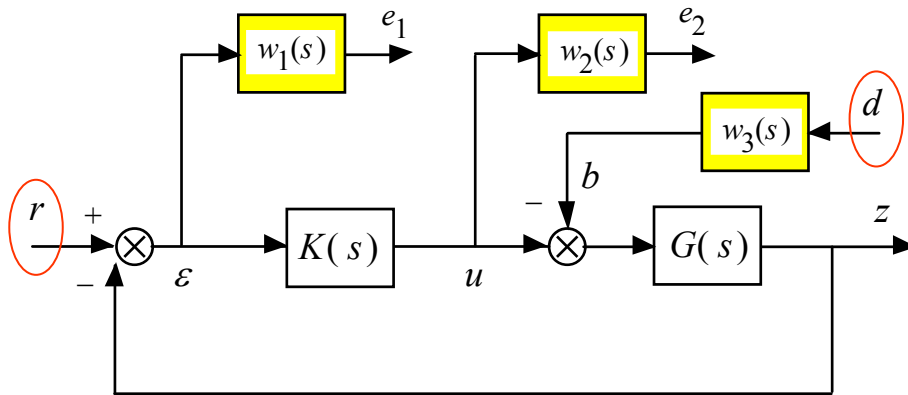
- K. Glover, J.C. Doyle. "State-Space Formulae for all Stabilizing Controllers That Satisfy an H_∞ -Norm Bound and Relations to Risk Sensitivity". *Systems & Control Letters*, vol. 11, pp. 167-172, 1988.
- J.C. Doyle, K. Glover, P.K. Khargonekar, B.A. Francis. "State-Space Solutions to Standard H_2 and H_∞ Control Problems". *IEEE Trans. Autom. Control*, AC 34 n° 8, pp. 831-846, 1989.
- P. Gahinet, P. Apkarian. "A Linear Matrix Inequality Approach to H_∞ Control". *Int. J. of Robust & Nonlinear Contr.*, vol. 4, pp. 421-448, 1994.
- T. Iwasaki, R.E. Skelton. "All Controllers for the General H_∞ Control Problem: LMI Existence Conditions and State-Space Formulas". *Automatica*, vol. 30 n° 8, pp. 1307-1317, 1994.

H_∞ – Typical 4 block problem

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H_∞ – Typical 4 block problem

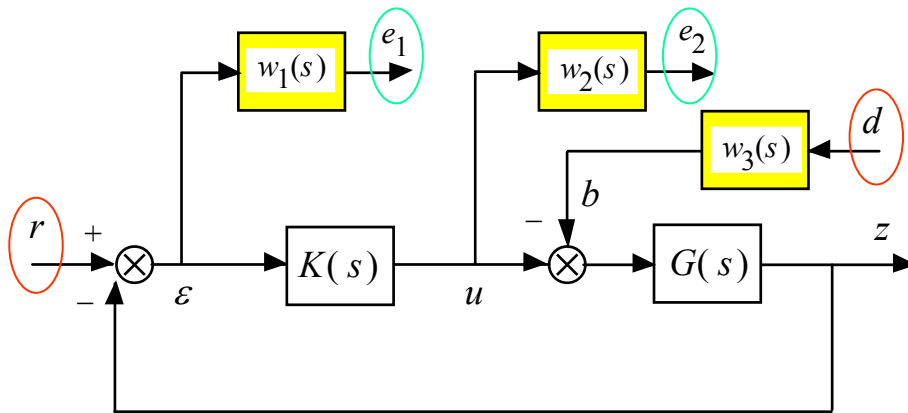


- 4 block criterion: two input / two output criterion

$$\begin{pmatrix} E_1(s) \\ E_2(s) \end{pmatrix} = \begin{pmatrix} w_1(s)S(s) & w_1(s)S(s)G(s)w_3(s) \\ w_2(s)K(s)S(s) & w_2(s)K(s)S(s)G(s)w_3(s) \end{pmatrix} \begin{pmatrix} R(s) \\ D(s) \end{pmatrix}$$

$$S = (I + GK)^{-1}$$

H_∞ – Typical 4 block problem

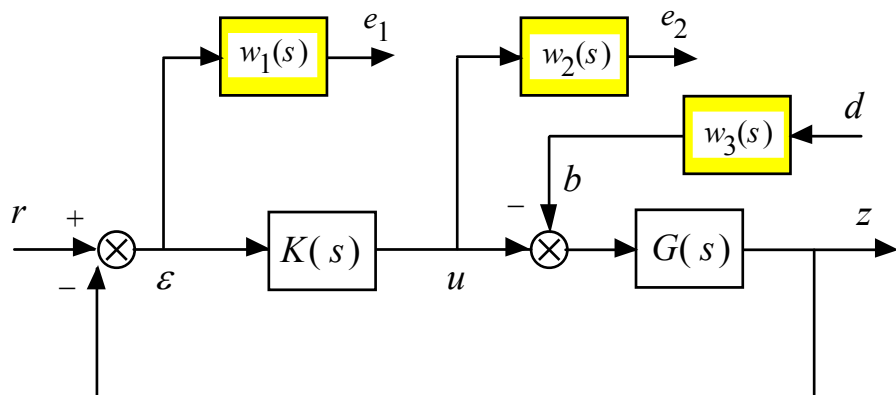


- 4 block problem: two input / two output criterion

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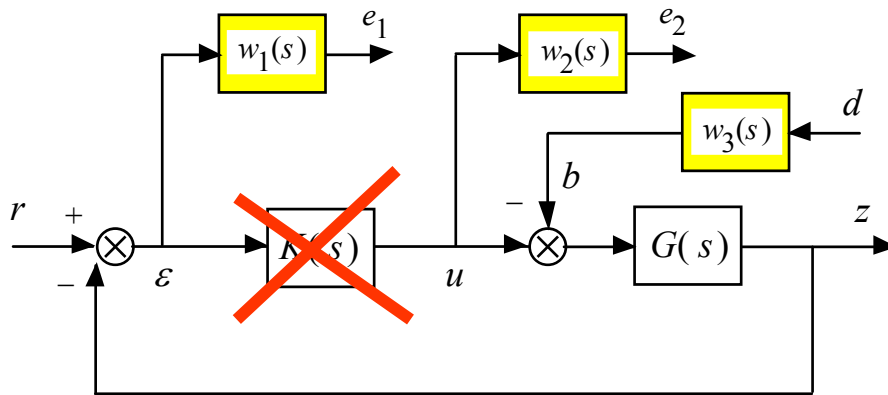


- 4 block problem: two input / two output criterion
- W_i : degrees of freedom

$$\begin{pmatrix} E_1(s) \\ E_2(s) \end{pmatrix} = \begin{pmatrix} w_1(s) S(s) & w_1(s) S(s) G(s) w_3(s) \\ w_2(s) K(s) S(s) & w_2(s) K(s) S(s) G(s) w_3(s) \end{pmatrix} \begin{pmatrix} R(s) \\ D(s) \end{pmatrix}$$

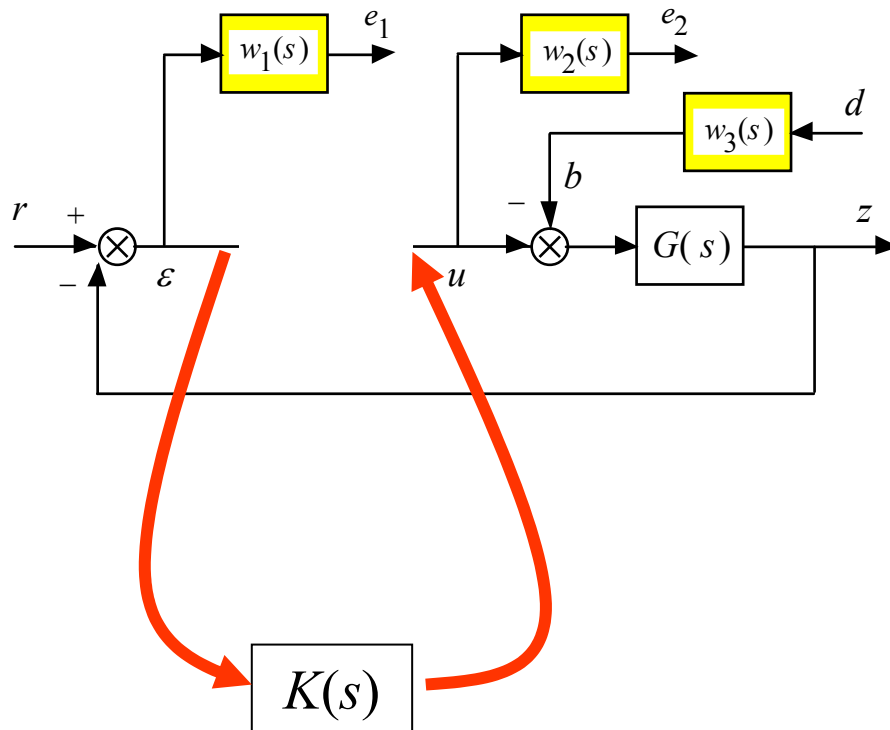
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H_∞ – Typical 4 block problem



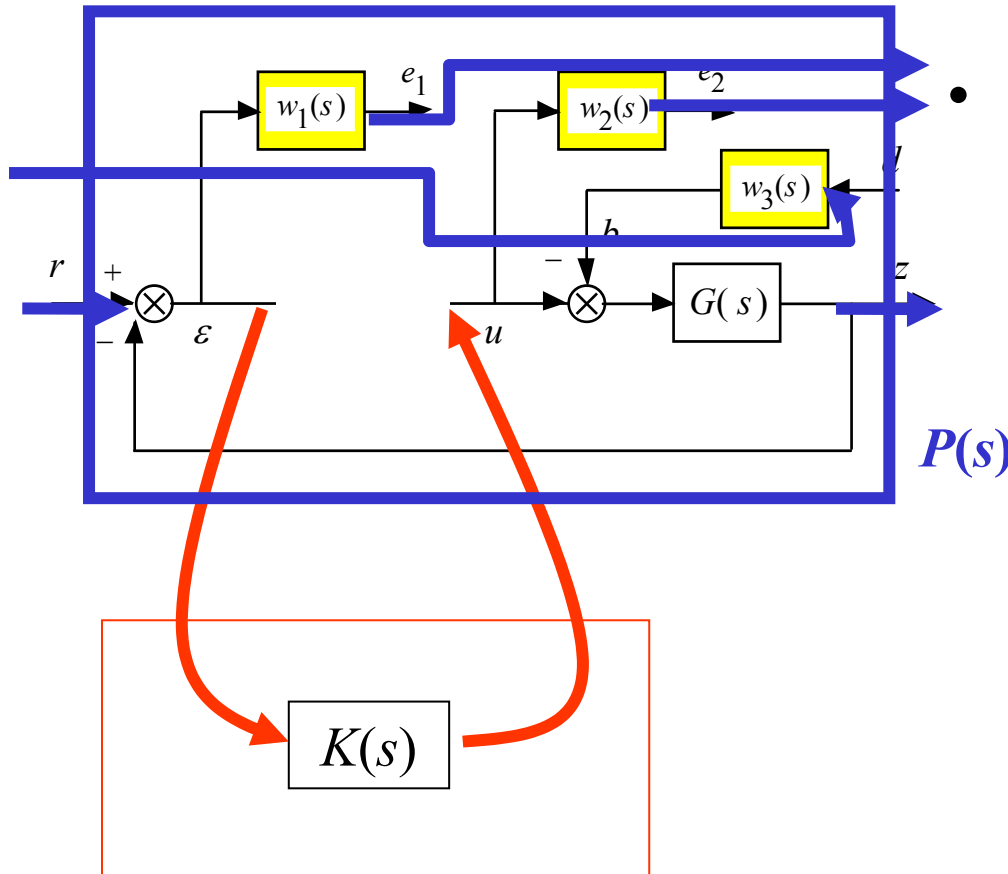
- Standard form
 - Isolate $K(s)$

H_∞ – Typical 4 block problem



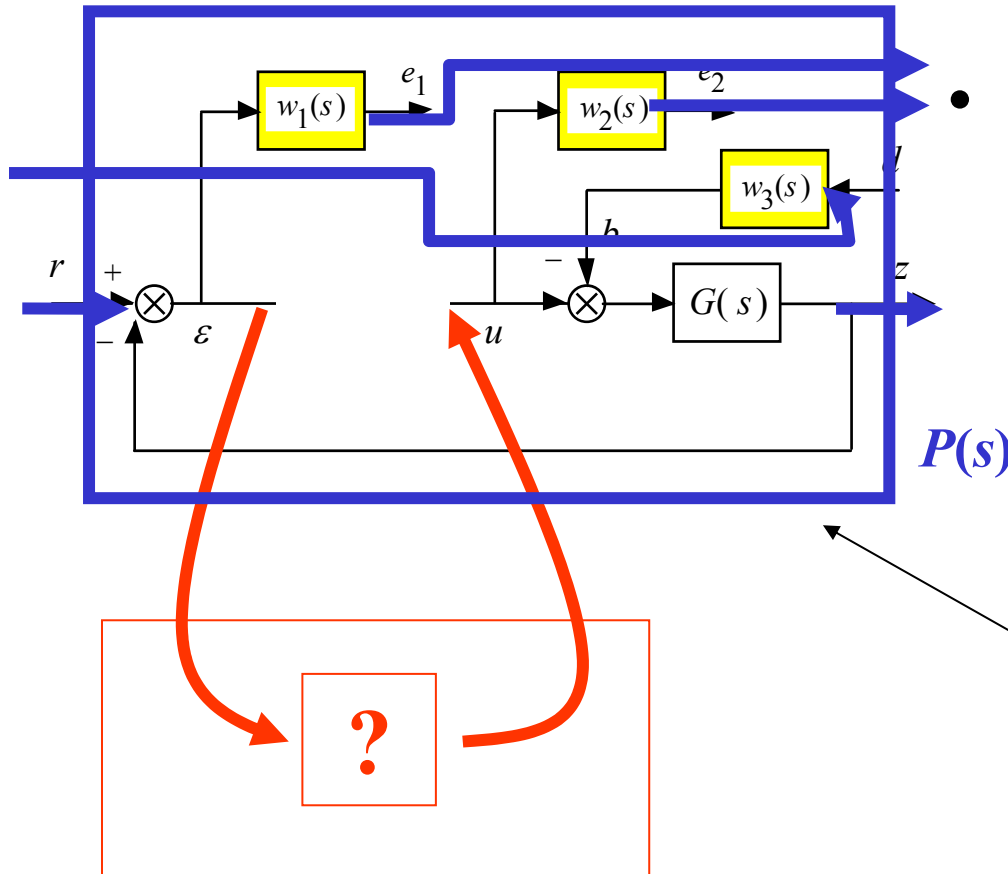
- Standard form
 - Isolate $K(s)$

H_∞ – Typical 4 block problem



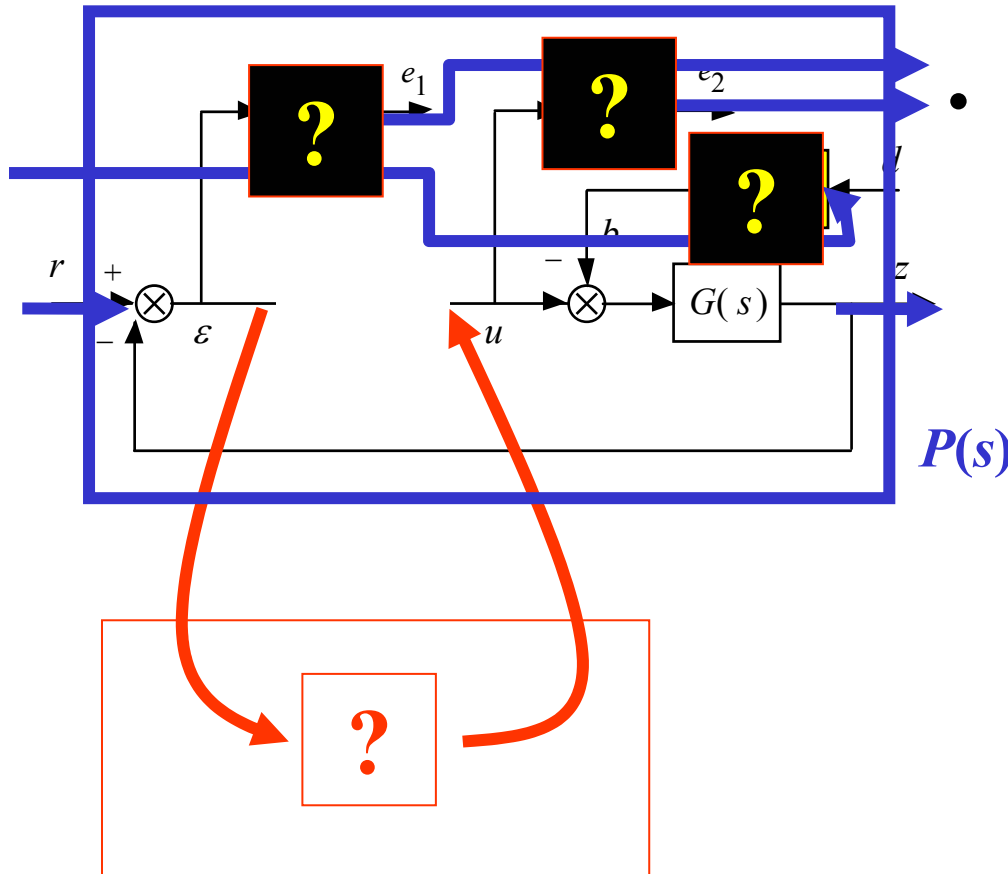
- Standard form
 - Isolate $K(s)$
 - split the graph (or the equations) in two parts

H_∞ – Typical 4 block problem



- Standard form
 - isolate $K(s)$
 - split the graph (or the equations) in two parts
 - obtain the augmented plan $P(s)$

H_∞ – Typical 4 block problem

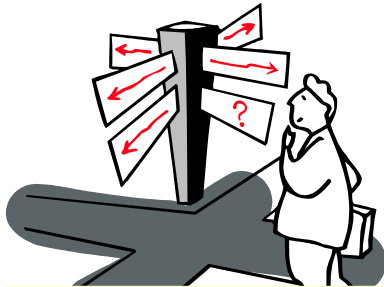


- Standard form
 - isolate $K(s)$
 - split the graph (or the equations) in two parts
 - obtain the augmented plant $P(s)$
 - **How to choose the weights W_i**

H_∞ – Methodology

Considered problem

What can we do with the
NP engineer questions?



Immersion
Multiplier
Relaxation



Convex optimisation



Global minima

are not mutually exclusive

H_∞ – Methodology

How to choose the weighting functions?

Block properties of the induced norm

$$\left\| \begin{pmatrix} L & M \\ N & O \end{pmatrix} \right\|_\infty < \gamma \Rightarrow \left\{ \begin{array}{l} \|L\|_\infty < \gamma \\ \|M\|_\infty < \gamma \\ \|N\|_\infty < \gamma \\ \|O\|_\infty < \gamma \\ \left\| \begin{pmatrix} L \\ N \end{pmatrix} \right\|_\infty < \gamma \\ \left\| \begin{pmatrix} M \\ O \end{pmatrix} \right\|_\infty < \gamma \\ \|(L \ M)\|_\infty < \gamma \\ \|(N \ O)\|_\infty < \gamma \end{array} \right.$$

H_∞ – Methodology

Application of this property to the 4 block criterion $\left\| \begin{pmatrix} w_1 S & w_1 S G w_3 \\ w_2 K S & w_2 K S G w_3 \end{pmatrix} \right\|_\infty < \gamma$

gives:

$$\|w_1 S\|_\infty < \gamma$$

$$\|w_2 K S\|_\infty < \gamma$$

$$\|w_1 w_3 S G\|_\infty < \gamma$$

$$\|w_2 w_3 K S G\|_\infty < \gamma$$

H_∞ – Methodology

Application of this property to the 4 block criterion $\left\| \begin{pmatrix} w_1 S & w_1 S G w_3 \\ w_2 K S & w_2 K S G w_3 \end{pmatrix} \right\|_\infty < \gamma$

gives: H_∞ norm property

$$\|w_1 S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega)| < \frac{\gamma}{|w_1(j\omega)|}$$

$$\|w_2 K S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega)| < \frac{\gamma}{|w_2(j\omega)|}$$

$$\|w_1 w_3 S G\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega) G(j\omega)| < \frac{\gamma}{|w_1(j\omega) w_3(j\omega)|}$$

$$\|w_2 w_3 K S G\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega) G(j\omega)| < \frac{\gamma}{|w_2(j\omega) w_3(j\omega)|}$$

H_∞ – Methodology

Application of this property to the 4 block criterion $\left\| \begin{pmatrix} w_1 S & w_1 S G w_3 \\ w_2 K S & w_2 K S G w_3 \end{pmatrix} \right\|_\infty < \gamma$

gives:

Closed loop function

$$\|w_1 S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega)| < \frac{\gamma}{|w_1(j\omega)|}$$

$$\|w_2 K S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega)| < \frac{\gamma}{|w_2(j\omega)|}$$

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H_∞ – Methodology

Application of this property to the 4 blocks criterion $\left\| \begin{pmatrix} w_1 S & w_1 S G w_3 \\ w_2 K S & w_2 K S G w_3 \end{pmatrix} \right\|_\infty < \gamma$

gives:

Closed loop function

$$\|w_1 S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega)| < \frac{\gamma}{|w_1(j\omega)|} \quad \text{Frequency constraints}$$

$$\|w_2 K S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega)| < \frac{\gamma}{|w_2(j\omega)|}$$

$$\|w_1 w_3 S G\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega) G(j\omega)| < \frac{\gamma}{|w_1(j\omega) w_3(j\omega)|}$$

$$\|w_2 w_3 K S G\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega) G(j\omega)| < \frac{\gamma}{|w_2(j\omega) w_3(j\omega)|}$$

H_∞ – Methodology

Shaping properties

Closed loop function

$$\|w_1 S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |S(j\omega)| < \frac{\gamma}{|w_1(j\omega)|}$$

Frequency constraints

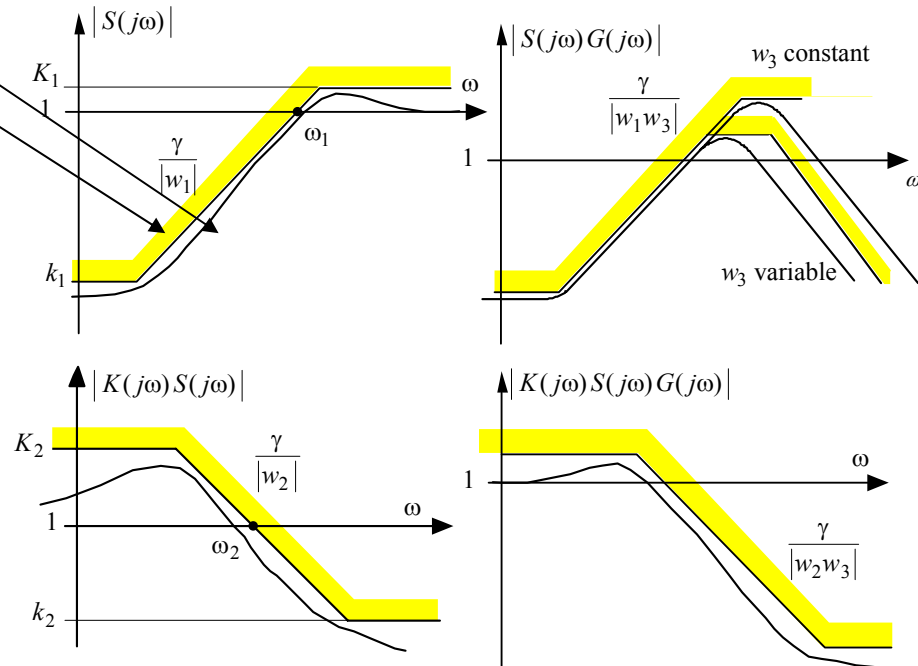
$$\|w_2 K S\|_\infty < \gamma \iff \forall \omega \in \mathbf{R} \quad |K(j\omega) S(j\omega)| < \frac{\gamma}{|w_2(j\omega)|}$$

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H_∞ – Methodology

Shaping properties
on closed loop
transfer functions



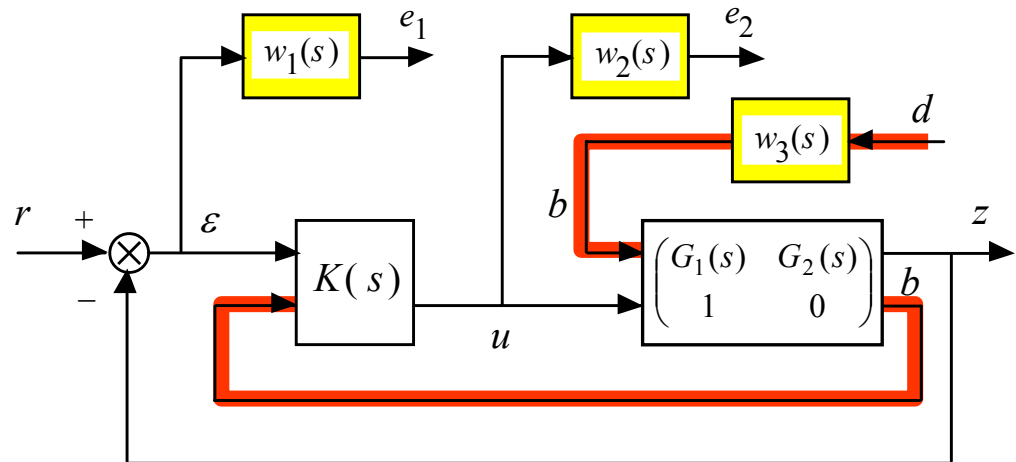
H_∞ – Methodology

- How to choose the weighting functions?
 - Use the shaping property
 - A game theory interpretation can be used too (worst case noise)
 - More generally, properties of the **induced norm** offer interesting possibilities for methodology and applications.
 - Many properties of induced norm are meaningful both from physical and mathematical point of view.

H_∞ – Flexibility of standard form

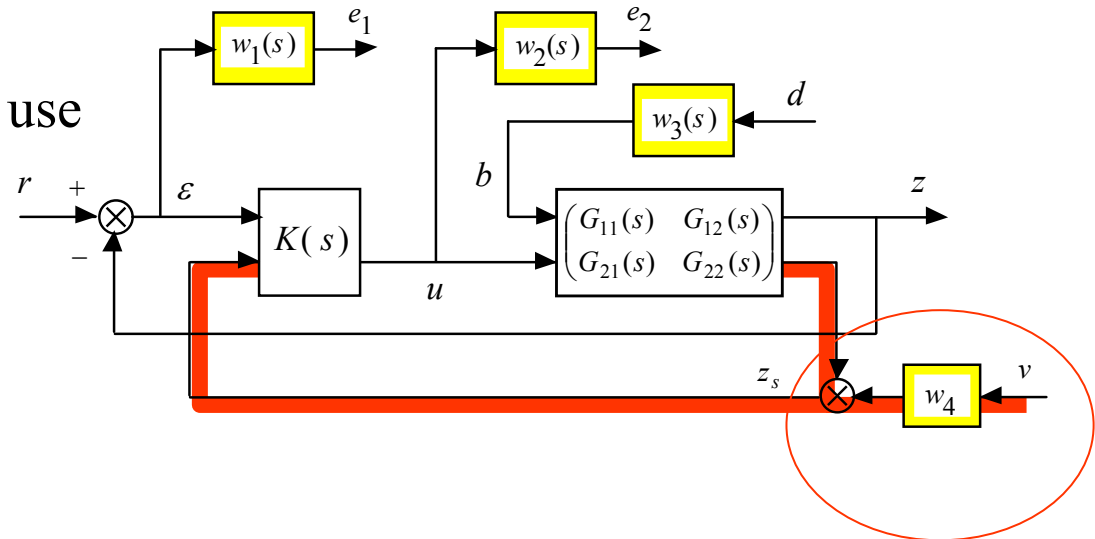
H_∞ – Flexibility of standard form

- The standard form of the problem allows to deal with many different problem in a unique framework
- Measurable perturbation rejection



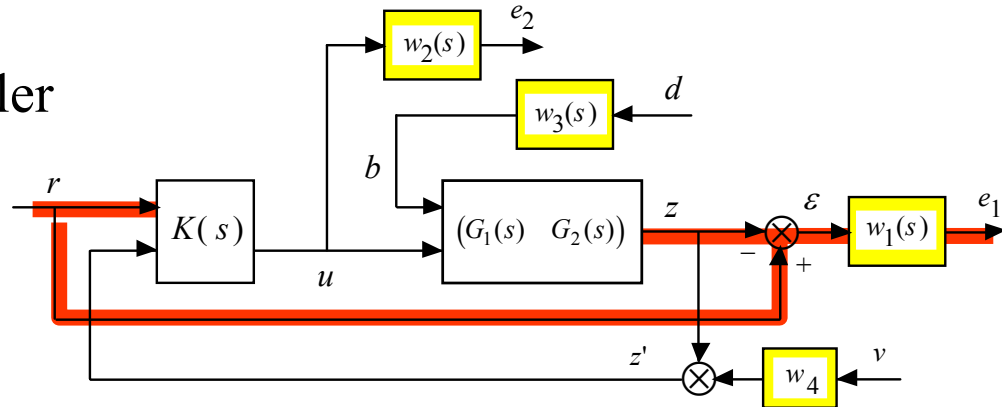
H_∞ – Flexibility of standard form

- The standard form of the problem allows to deal with many different problem in the unique framework
- Extra measurement use



H_∞ – Flexibility of standard form

- The standard form of the problem allows to deal with many different problem in the unique framework
- Two degrees of freedom controller



H_∞ – Applications

H_∞ – Applications

Since 1990, the methodology had been largely developed and this approach is now currently used

This method is applied in many domains, to cite a few

- Dam regulation (Supélec-EDF, INRA-Cémagref)
- Electromechanical drive (Supélec-CRAN)
- Irrigation canal control (INRA-Cémagref-LAP)
- Nuclear power plant (EDF > 2 Phd)
- Hydraulic power plant (Supélec-EDF)
- Missile control (Supélec-Aerospatiale > 4 Phd)
- Launch vehicle (Supélec-CNES > 2 Phd)
- Freezing control process (INRA-Cémagref)
- Magnetic Bearing (Supélec-S2M)
- ...

and many places (research departments and industries)

H_∞ – Summary

H_∞ – Summary

- **Standard form** is very powerful (idea of graph separation)
- Properties of the **induced norm** offer many interesting possibilities
- Induced norm properties are closely related to the idea of energy, amplification. Other approach using game theory offer other methodology.
- With an adapted methodology, this method can deal with performance and robustness specs in the same time with an **input/output point of view** (it allows to guarantee multivariable stability margin in the output feedback case!).

μ analysis

μ – Definition

μ – Definition

- Def. Structured set of perturbations:

$$\underline{\Delta} := \left\{ \begin{array}{l} \Delta = \text{diag} \left\{ \Delta_1, \dots, \Delta_q, \delta_1 I_{r_1}, \dots, \delta_r I_{r_r}, \varepsilon_1 I_{c_1}, \dots, \varepsilon_c I_{c_c} \right\} \in \mathbf{C}^{k \times k} \\ \Delta_i \in \mathbf{C}^{k_i \times k_i} \quad ; \quad \delta_i \in \mathbf{R} \quad ; \quad \varepsilon_i \in \mathbf{C} \end{array} \right\}$$

- Def. Structured singular value of P for the structure $\underline{\Delta}$:

$$\mu_{\underline{\Delta}}(P) := \left(\inf_{\Delta \in \underline{\Delta}} \left(\overline{\sigma}(\Delta) : \det(I - \Delta P) = 0 \right) \right)^{-1}$$

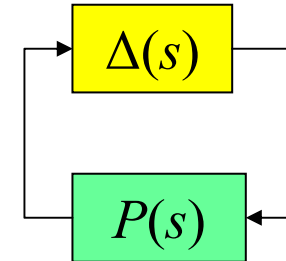
$$\mu_{\underline{\Delta}}(P) := 0 \quad \text{si} \quad \forall \Delta \in \underline{\Delta} \quad \det(I - \Delta P) \neq 0$$

- If $\mu \neq 0$, μ^{-1} is the size of the smallest Δ that makes $I - \Delta P$ rank deficient.

μ – Definition – Property

- *Def. Structured operator $\Delta(s)$*

$$\forall \omega \in \mathfrak{R} \quad \Delta(j\omega) \in \underline{\Delta}$$



- *Fundamental property*

- *$P(s)$ stable*

- *Connexion $P(s)*\Delta(s)$ is internally stable for all structured $\Delta(s)$ stable and $\|\Delta(s)\|_\infty < \alpha$*

if and only if

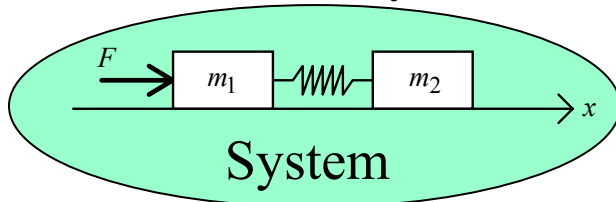
$$\forall \omega \quad \mu_{\underline{\Delta}}(P(j\omega)) \leq \alpha^{-1}$$

μ – Definition

- J.C. Doyle. "Structured Uncertainty in Control Systems Design". 24th Conf. on Decision and Control, pp. 260-265, Ft Lauderdale, Floride, 1985.
- K. Zhou, J.C. Doyle, K. Glover, Robust and Optimal Control. Prentice-Hall, 1996.

μ – basic example

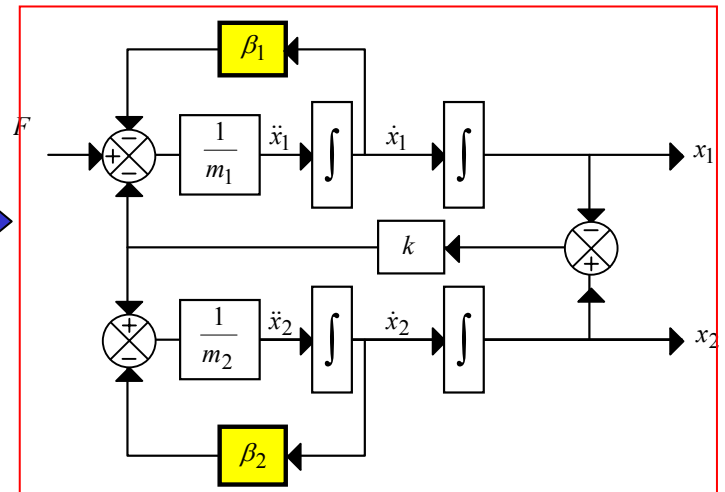
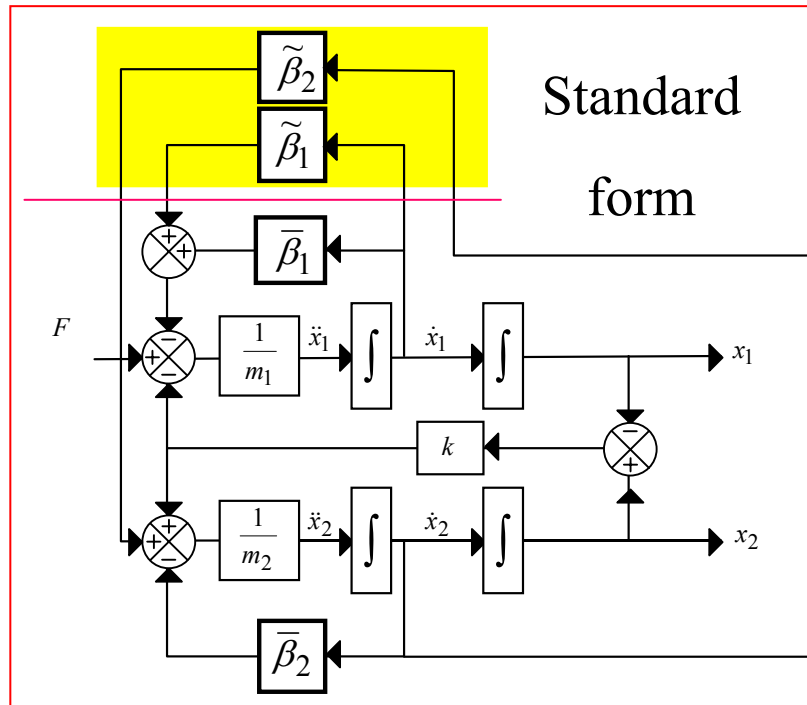
μ – basic example



Model

$$\begin{cases} m_1 \ddot{x}_1 = F - k(x_2 - x_1) - \beta_1 \dot{x}_1 \\ m_2 \ddot{x}_2 = +k(x_2 - x_1) - \beta_2 \dot{x}_2 \end{cases}$$

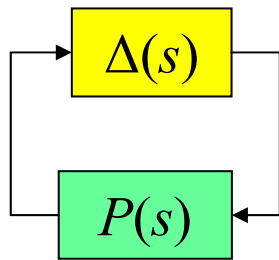
+ uncertainties on β_1 and β_2



μ –uncertainty interpretation

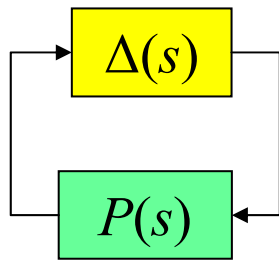
- In general, uncertainties stand for
 - Unknown bounded parameters
 - Imperfect measurements (high frequencies ...)
 - Simplification of equations (intentional ... or not)
 - ...

μ – standard form



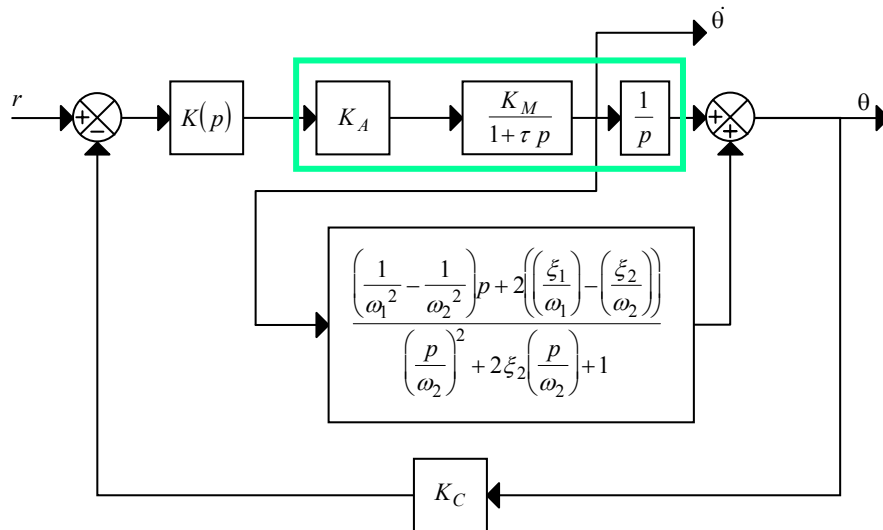
- The standard form allows to separate
 - the uncertainties
 - known part of the model

μ – standard form



- The standard form allows to separate
 - the uncertainties
 - known part of the model
- Is this approach universal?
 - Can we compute the standard form in more complex situations ?
 - Technical and long answer
 - ...
 - Let see an example

μ – standard form



- Electrical motor with flexible mode

$$K_M \in [10 \times 0.75, 10 \times 1.25]$$

$$\tau \in [0.015 \times 0.75, 0.015 \times 1.25]$$

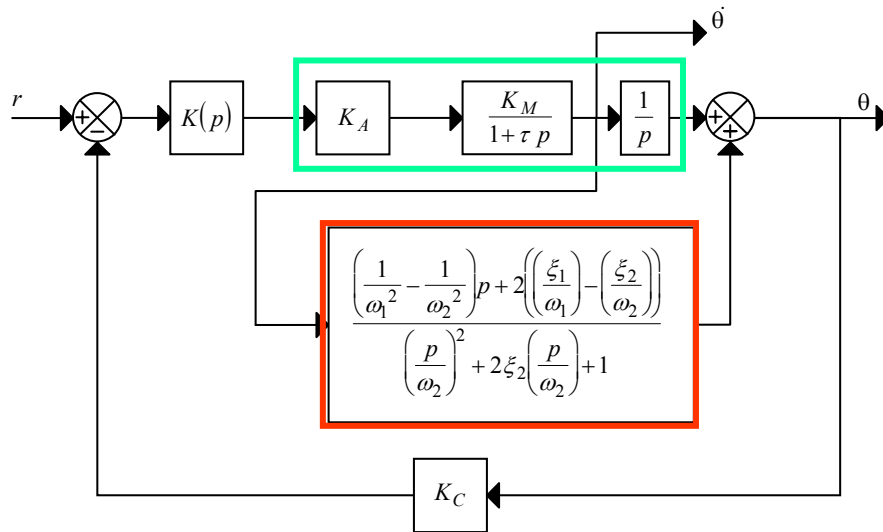
$$\xi_1 \in [0.02 \times 0.75, 0.02 \times 1.25]$$

$$\xi_2 \in [0.01 \times 0.75, 0.01 \times 1.25]$$

$$\omega_1 \in [200 \times (1 - \beta), 200 \times (1 + \beta)]$$

$$\omega_2 \in [300 \times (1 - \beta), 300 \times (1 + \beta)] \quad \beta = (1 - 0.88456) \times 0.5$$

μ – standard form



- Electrical motor with flexible modes

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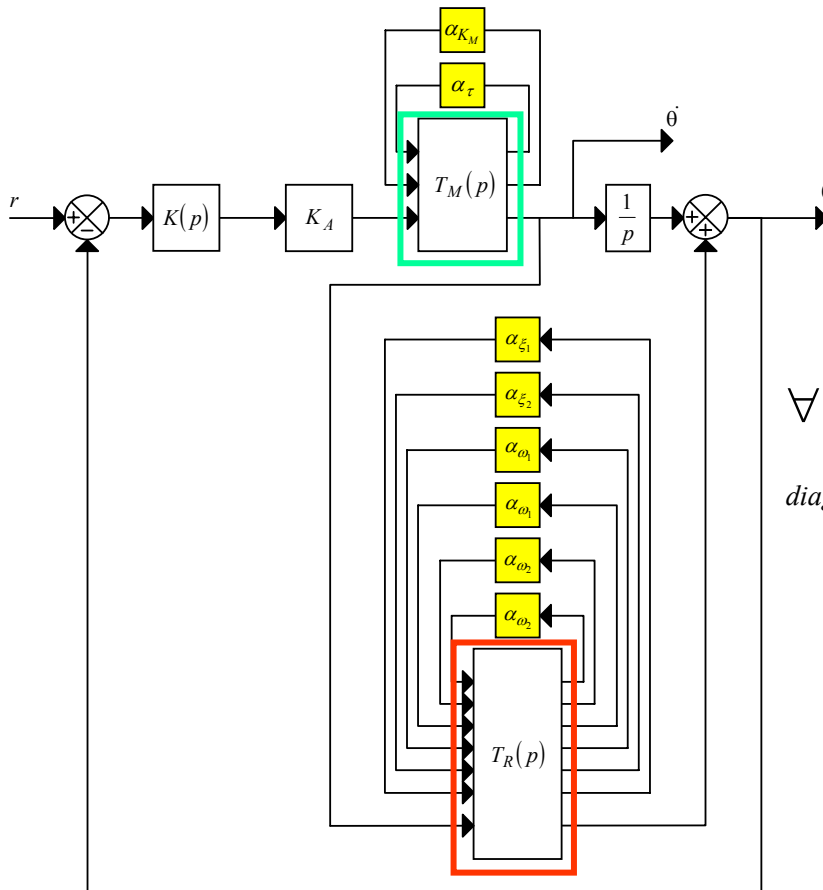
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μ – standard form



- We want to separate unknown parameters to check the fundamental property:

$$\forall \omega \quad \mu_{\text{diag}(\mathbf{R}, \mathbf{R}, \mathbf{R}, \mathbf{R}, \mathbf{R}I_2, \mathbf{R}I_2)}(T_\mu(i\omega)) < 1$$

$$\text{diag}(\alpha_{K_M}, \alpha_\tau, \alpha_{\xi_1}, \alpha_{\xi_2}, \alpha_{\omega_1}, \alpha_{\omega_1}, \alpha_{\omega_2}, \alpha_{\omega_2}) \in \text{diag}(\mathbf{R}, \mathbf{R}, \mathbf{R}, \mathbf{R}, \mathbf{R}I_2, \mathbf{R}I_2)$$

- We have to compute the state space form of T_M and T_R

μ – standard form

- For the central part of the motor model, we obtain the following result (symbolic computation approach)

$$T_M(p) = \left(\begin{array}{c|cc} -\frac{1}{\tau_0} & \begin{bmatrix} -K_{M1} & -\frac{\tau_1}{\tau_0} & -K_{M0} \end{bmatrix} \\ \hline \begin{bmatrix} 0 \\ 1 \\ -\frac{1}{\tau_0} \\ -\frac{1}{\tau_0} \end{bmatrix} & \begin{bmatrix} 0 & 0 & 1 \\ 0 & -\frac{\tau_1}{\tau_0} & 0 \\ 0 & -\frac{\tau_1}{\tau_0} & 0 \end{bmatrix} \end{array} \right)$$

$$\begin{cases} K_{M0} = 42 \\ K_{M1} = 42 \times 0.25 \\ \tau_0 = 0.015 \\ \tau_1 = 0.015 \times 0.25 \end{cases}$$

μ – standard form

- And for the flexible mode we obtain

$$T_R(p) = \left(\begin{array}{c|c} \left[\begin{array}{cccc} 0 & -2\xi_{10}\omega_{20}^2 & 0 & 0 \\ 0 & -2\xi_{20}\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}^2} & -\frac{2\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right] & \left[\begin{array}{ccccccc} \xi_{11} & 0 & 0 & 0 & \xi_{10}\omega_{21} & \xi_{10}\omega_{20}\omega_{21} & \xi_{10}\omega_{20}^2 \\ 0 & \xi_{21} & 0 & 0 & 0 & \xi_{20}\omega_{21} & \xi_{20}\omega_{20} \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} & -\frac{\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}\omega_{21}}{\omega_{10}^2} & -\frac{\omega_{10}^2 + \omega_{20}^2}{\omega_{10}^2} \\ 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] \\ \hline \left[\begin{array}{cccc} 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}} & -2\frac{\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\frac{\omega_{20}^2}{\omega_{10}} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] & \left[\begin{array}{ccccccc} 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 \\ 0 & 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20} \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} & \frac{\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}^2}{\omega_{10}^2} \\ 0 & 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & \frac{\omega_{21}}{\omega_{10}} & \frac{\omega_{20}\omega_{21}}{\omega_{10}} & \frac{\omega_{20}^2}{\omega_{10}} \\ 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 & \omega_{20} \\ 0 & 0 & 0 & 0 & \omega_{10} & \omega_{10} & \omega_{10} \\ 0 & 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20} \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] \end{array} \right)$$

$$\left\{ \begin{array}{l} \xi_{10} = 0.02 \\ \xi_{11} = 0.02 \times 0.25 \\ \omega_{10} = 200 \\ \omega_{11} = 200 \times (1 - 0.88456) \times 0.5 \end{array} \right. \quad \left\{ \begin{array}{l} \xi_{20} = 0.01 \\ \xi_{21} = 0.01 \times 0.25 \\ \omega_{20} = 300 \\ \omega_{21} = 300 \times (1 - 0.88456) \times 0.5 \end{array} \right.$$

μ – standard form

- And for the flexible mode we obtain
- Don't make it manually!

$$T_R(p) = \left(\begin{array}{c|c} \left[\begin{array}{cccc} 0 & -2\xi_{10}\omega_{20}^2 & 0 & 0 \\ 0 & -2\xi_{20}\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}^2} & -\frac{2\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right] & \left[\begin{array}{ccccccc} \xi_{11} & 0 & 0 & 0 & \xi_{10}\omega_{21} & \xi_{10}\omega_{20}\omega_{21} & \xi_{10}\omega_{20}^2 \\ 0 & \xi_{21} & 0 & 0 & 0 & \xi_{20}\omega_{21} & \xi_{20}\omega_{20} \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} & -\frac{\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}\omega_{21}}{\omega_{10}^2} & -\frac{\omega_{10}^2 + \omega_{20}^2}{\omega_{10}^2} \\ 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] \\ \hline \left[\begin{array}{cccc} 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}} & -2\frac{\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\frac{\omega_{20}^2}{\omega_{10}} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] & \left[\begin{array}{ccccccc} 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 \\ 0 & 0 & 0 & 0 & 0 & \omega_{21} & \omega_{20} \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} & \frac{\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}\omega_{21}}{\omega_{10}^2} & \frac{\omega_{20}^2}{\omega_{10}^2} \\ 0 & 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & \frac{\omega_{21}}{\omega_{10}} & \frac{\omega_{20}\omega_{21}}{\omega_{10}} & \frac{\omega_{20}^2}{\omega_{10}} \\ 0 & 0 & 0 & \omega_{21} & \omega_{20}\omega_{21} & \omega_{20}^2 & \omega_{20} \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] \end{array} \right)$$

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μ – standard form

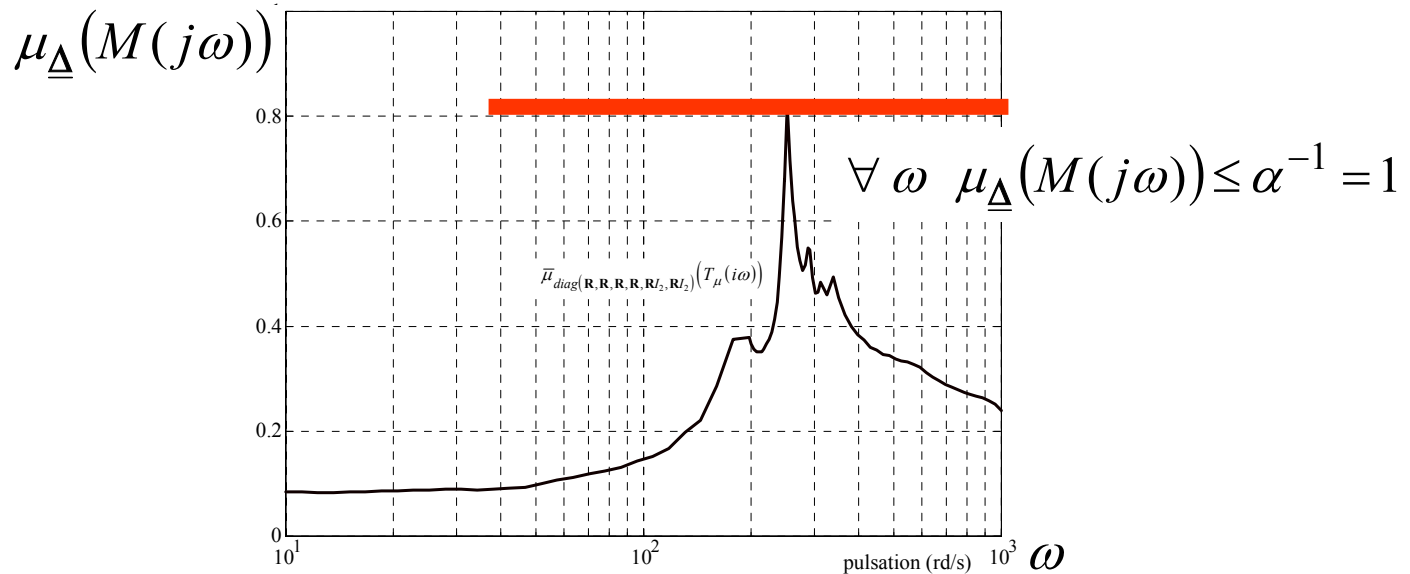
- And for the flexible mode we obtain
- Don't make it by hand !
- *A free toolbox is available (numerical approach using Matlab):*
 - *J. F. Magny, CERT, Toulouse.*

$$T_R(p) = \left(\begin{array}{c|c} \left[\begin{array}{cccc} 0 & -2\xi_{10}\omega_{20}^2 & 0 & 0 \\ 0 & -2\xi_{20}\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}^2} & -\frac{2\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right] & \left[\begin{array}{cccc} \xi_{11} & 0 & 0 & 0 \\ 0 & \xi_{21} & 0 & 0 \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} \\ 0 & 0 & 0 & 0 \end{array} \right] \\ \hline \left[\begin{array}{cccc} 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ \frac{2}{\omega_{10}} & -2\frac{\omega_{20}^2}{\omega_{10}^2} & 0 & 0 \\ 0 & -2\frac{\omega_{20}^2}{\omega_{10}} & 0 & 0 \\ 0 & -2\omega_{20}^2 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] & \left[\begin{array}{cccc} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -\frac{\omega_{11}}{\omega_{10}} & -\frac{\omega_{11}}{\omega_{10}^2} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \end{array} \right)$$

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μ – standard form

- So we can check the property :



- Stability is obtained for all models in the defined set

μ – standard form

- The standard form allows to separate
 - deterministic and known part of the model
 - from the uncertainties
- Is this approach universal?
 - **Yes** (in fact *almost* all the time)
 - Using adapted tools, standard form can be easily computed

μ – computation

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- Direct computation
 - The value of μ can be approximated by gridding method for small dimension of the problem

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 - Exact computation is possible for some degenerated situations (for some *very small problem*, upper bound of μ is equal to μ , and so μ can be easily computed).

μ – computation

- Direct computation
 - The value of μ can be approximated by gridding method for small dimension of the problem
 - Exact computation is possible for some degenerated situations (for some small problem, upper bound of μ is equal to μ , and so μ can be easily computed).
 - In general, the exact computation is a **non polynomial** problem considering the number of uncertainties.
- Braatz, R.D., P.M. Young, J.C. Doyle and M. Morari.
"Computational Complexity of mu Calculation, IEEE Trans. *Autom. Control*, AC 39 n° 5, pp. 1000-1002, 1994.

μ – computation

- Indirect computation using properties of μ

$$P1 \quad \forall \underline{\Delta}, \forall \alpha \in \mathbf{C}, \forall P \in \mathbf{C}^{k \times k}, \mu_{\underline{\Delta}}(\alpha P) = |\alpha| \mu_{\underline{\Delta}}(P)$$

$$P2 \quad \underline{\Delta}_1 = \mathbf{C}^{k \times k} \quad \mu_{\underline{\Delta}_1}(P) = \overline{\sigma}(P)$$

$$P3 \quad \underline{\Delta}_2 = \{\delta I_k, \delta \in \mathbf{R}\} \quad \mu_{\underline{\Delta}_2}(P) = \rho_R(P) := \sup_{\lambda \in \mathbf{R}} (|\lambda| : \det(\lambda I_k - P) = 0)$$

$$P4 \quad \underline{\Delta}_3 = \{\delta I_k, \delta \in \mathbf{C}\} \quad \mu_{\underline{\Delta}_3}(P) = \rho(P) := \sup_{\lambda \in \mathbf{C}} (|\lambda| : \det(\lambda I_k - P) = 0)$$

$$P5 \quad \forall P \in \mathbf{C}^{k \times k} \quad \underline{\Delta}_a \subset \underline{\Delta}_b \Rightarrow \rho_{\underline{\Delta}_a}(P) \leq \mu_{\underline{\Delta}_a}(P) \leq \mu_{\underline{\Delta}_b}(P) \leq \overline{\sigma}(P)$$

$$P6 \quad \forall P \in \mathbf{C}^{k \times k}, \forall D \in \mathbf{D}, \mu_{\underline{\Delta}}(P) \leq \overline{\sigma}(D P D^{-1})$$

$$P7 \quad \forall P \in \mathbf{C}^{k \times k}, \forall U \in \mathbf{U}, \rho_R(U P) \leq \mu_{\underline{\Delta}}(P)$$

$$P8 \quad \forall P \in \mathbf{C}^{k \times k}, Q \in \mathbf{Q}, \rho_R(Q P) \leq \mu_{\underline{\Delta}}(P)$$

μ – computation

- Indirect computation using properties of μ

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$$P8 \quad \forall P \in \mathbf{C}^{k \times k}, Q \in \mathbf{Q}, \rho_R(Q P) \leq \mu_{\underline{\Delta}}(P)$$

μ – computation

- *Upper bound*

$$P6 \Rightarrow \forall P \in \mathbf{C}^{k \times k}, \mu_{\underline{\Delta}}(P) \leq \min_{D \in \mathbf{D}} \bar{\sigma}(D P D^{-1})$$

convex problem

D hermitian, $D \geq 0$, $D\Delta = \Delta D$

- *LMI formulation*

$$\exists D \in \mathbf{D}: \bar{\sigma}(D P D^{-1}) \leq \alpha$$

$$\exists D \in \mathbf{D}: \bar{\lambda}(D^{-1} P^* D D P D^{-1}) \leq \alpha^2$$

$$\exists D \in \mathbf{D}: D^{-1} P^* D^2 P D^{-1} - \alpha^2 I \leq 0$$

$$\exists D \in \mathbf{D}: P^* D^2 P - \alpha^2 D^2 \leq 0$$

$$\exists D' \in \mathbf{D}: P^* D' P - \alpha^2 D' \leq 0$$

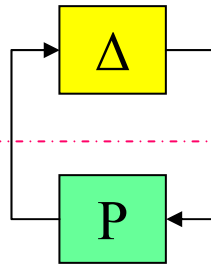
$$\alpha^* = \min_{D \in \mathbf{D}} \alpha$$

$$\alpha \geq 0$$

$$P^* D P - \alpha^2 D \leq 0$$

μ – graph separation approach

μ – graph separation approach

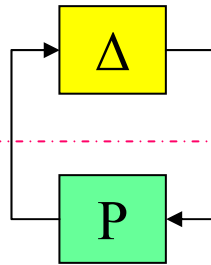


Does the closed loop have a
good property P_{cl} ?



Use properties of each
operator (P_a, P_b) that
guarantee a global property
of the closed loop P_{cl}

μ – graph separation approach



Do the closed loop have a good property P_{cl} ?



Use properties of each operator (P_a, P_b) that guarantee a global property of the close loop P_{cl}

- If P_{cl} is “closed loop stability property”
 - Sufficient condition (small gain theorem):

$$P_a : \|\Delta\| < 1 \text{ et } P_b : \|P\| \leq 1$$

- Sufficient condition (passivity condition):

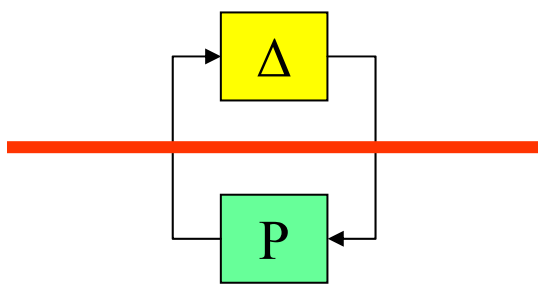
$$\exists \varepsilon, \forall t > 0, \forall x, \|x\| < 1$$

$$P_a : \int_0^t x^T H_1 x > \varepsilon > 0 \text{ et } P_b : \int_0^t x^T H_2 x < -\varepsilon < 0$$

- others ...

μ – graph separation approach

- Application to the structured singular value
 - The property P_{cl} is: rank deficiency of $I - \Delta P$



$$P_a : \bar{\sigma}(\Delta) < \alpha^{-1} \quad \text{and} \quad P_b : \bar{\sigma}(P) = \alpha$$

⇓

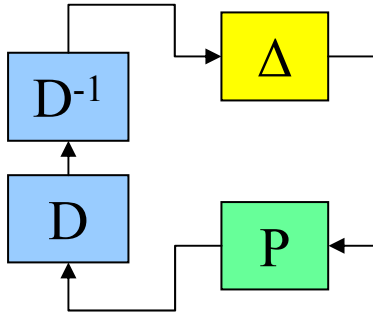
$$P_{cl} : I - \Delta P \text{ not rank deficient}$$

- With the mu formalism : $\mu_{\underline{\Delta}}(P) \leq \alpha = \bar{\sigma}(P)$
- (There is no rank deficiency in the ball whose diameter is α)
- *Note: we can obtain this result using properties*

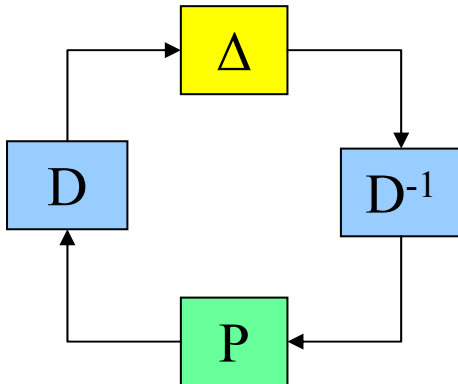
$$\underbrace{\Delta \subset \mathbf{C}^{k \times k} \Rightarrow \mu_{\underline{\Delta}}(P) \leq \underbrace{\mu_{\mathbf{C}^{k \times k}}(P) = \bar{\sigma}(P)}_{P2}}_{P5}$$

μ – graph separation approach

- We can use a multiplier D



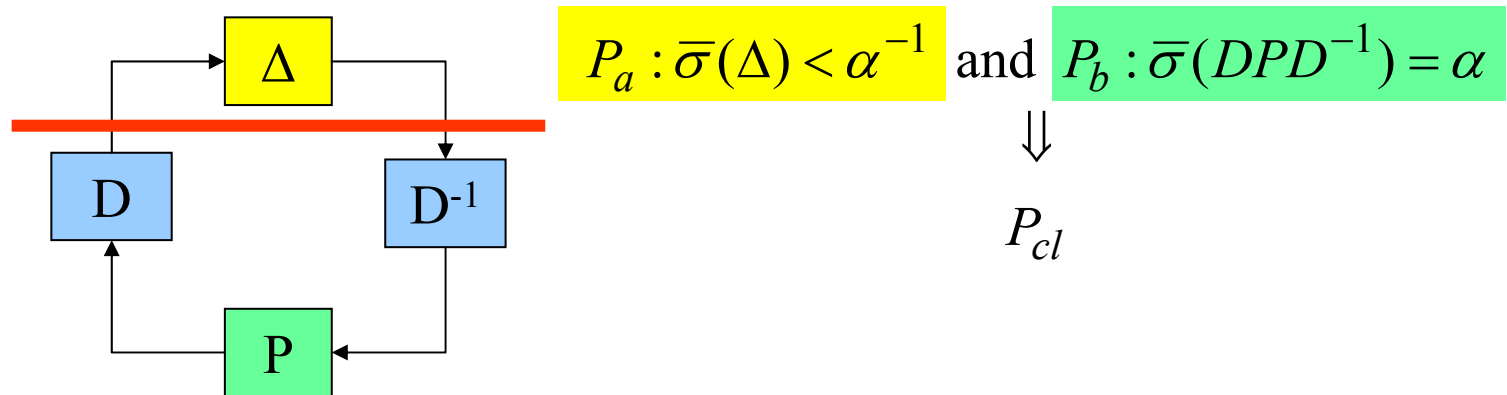
- And reorganized the operators



Conservation of
the property
 P_{cl} must be
checked

μ – graph separation approach

- Graph separation



- With the mu formalism : $\mu_{\underline{\Delta}}(P) \leq \alpha = \bar{\sigma}(DPD^{-1})$
- We get a less conservatism result

Note: we find out the property P6.

μ – graph separation approach

- This graph separation approach can be **generalized to many classes of operators** (nonlinear memoryless functions, time varying systems, nonlinear systems)
- Safonov, M. G., Stability and Robustness of multivariable Feedback Systems, MIT Press, Cambridge, 1980.

Conclusions

Conclusion

- What is a good method in automatic control ?
- For:
 - Identification
 - Synthesis
 - analysis
- Looking for methods who are:
 - general
 - easy to use
 - CAD (computer-aided design)

Conclusion

- Since 1980, robustness theory had been developed in this state of mind (impressive development after 1992)
 - *General* criterion formulation
 - Specs are *formalised*
 - Parametric uncertainties are *considered in the design*
 - Neglected dynamics are *considered in the design*
 - Formulation as an optimisation problem is very general and allows a *CAD approach*

Conclusions

μ analysis and H_∞ share some fundamental concepts and objects :

- standard form**
- input/output approach**
- induced norms and associated properties**
- multiplier approach**
- possibility to reach tractable problems**

Those concepts are important and gather many ideas of post modern robust control

Our aim is to enlarge the field of robust methods to the nonlinear systems, while preserving:

- global properties for stability**
- global performance specs**
- direct consideration of robust demands (margin...)**

