

GRADUATE COURSE ON
POLYNOMIAL METHODS FOR
ROBUST CONTROL
PART III.2

**ROBUST DESIGN:
PARAMETRIZATION OF ALL
CONTROLLERS**

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Disks of Newton
František Kupka (1871-1957)

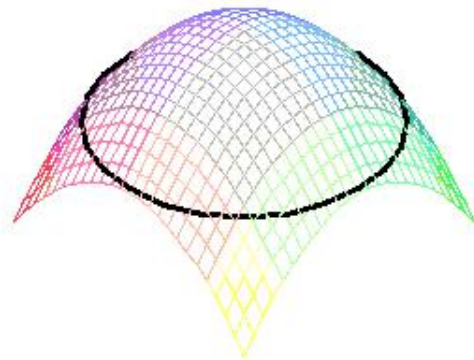
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Summary

We have seen that robust design can be performed by **pole placement**, provided a good **approximation** of the non-convex stability region of a fixed degree **polynomial**

Alternatively, **non-convexity** of the stability region can also be overcome by a **rational parametrization of all stabilizing controllers**

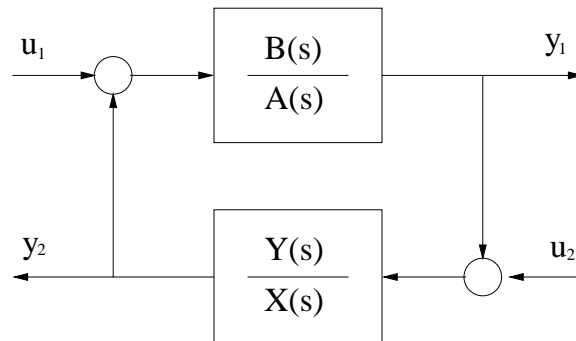
The parametrization is infinite-dimensional, but it turns out to be **convex** which is crucial for design



In the sequel we show that it allows to parametrize in a convex way the whole set of **robustly stabilizing controllers** for a specific class of uncertainty

Rational coprime factorization

We consider the controller-plant setup



where plant and controller transfer function

$$\frac{b(s)}{a(s)}, \quad \frac{y(s)}{x(s)}$$

are now ratios of coprime **proper stable rational** functions (and not polynomials anymore)

Example

Plant

$$\frac{b(s)}{a(s)} = \frac{s-1}{s+2}$$

can be written as

$$\frac{b(s)}{a(s)} = \frac{\left(\frac{s-1}{s+1}\right)}{\left(\frac{s+2}{s+1}\right)}$$

where both $a(s)$ and $b(s)$ are **proper stable rational**

Normalized coprime factorization

A suitable choice of a coprime factorization of a rational function

$$\frac{b(s)}{a(s)} = \frac{\left(\frac{b(s)}{c(s)}\right)}{\left(\frac{a(s)}{c(s)}\right)}$$

is the so-called **normalized** factorization, where polynomial $c(s)$ solves the quadratic **spectral factorization** equation

$$a^*(s)a(s) + b^*(s)b(s) = c^*(s)c(s)$$

In ct $c^*(s) = c(-s)$, in dt $c^*(z) = c(z^{-1})$

Example

Plant

$$\frac{b(s)}{a(s)} = \frac{s-1}{s+2}$$

Denominator polynomial $c(s)$ satisfies

$$\begin{aligned} (s-1)^*(s-1) + (s+2)^*(s+2) \\ = (-s-1)(s-1) + (-s+2)(s+2) \\ = 5 - s^2 = c^*(s)c(s) \end{aligned}$$

Solving the quadratic equation, we obtain the stable polynomial

$$c(s) = \sqrt{5} + \sqrt{2}s$$

Internal stability

It turns out that the whole set of proper stable rational functions (further denoted by R) is a **ring**, just as the set of polynomials

We will encounter the **Diophantine** equation or **Bézout** equation (from Etienne Bézout [1730 Nemours - 1783 Fontainebleau], no photo sorry)

$$ax + by = 1$$

now defined over R

A controller y/x **internally stabilizes** the plant b/a iff all the 4 transfer functions

$$\begin{aligned} z_1 &= \frac{bx}{ax + by} u_1 & z_1 &= \frac{-by}{ax + by} u_2 \\ z_2 &= \frac{-by}{ax + by} u_1 & z_2 &= \frac{-ay}{ax + by} u_2 \end{aligned}$$

are stable

We see that denominator $ax + by$ must be a **unit** in R (i.e. reciprocal also belongs to R)

Parametrization of all stabilizing controllers

Frequently called **Youla** or **Youla-Kučera** or also **Q**-parametrization, published around 1975

All stabilizing controllers are generated by all solution pairs to Bézout equation

$$\frac{y}{x} = \frac{y_0 - aq}{x_0 + bq}$$

where x_0, y_0 is a particular solution and q is any proper stable rational function such that $x_0 + bq \neq 0$

This fundamental result launched the entire area of research on **polynomial methods**

Design in **three steps**:

1. Express plant as ratio of elements in R
2. Find particular solution to Bézout equation
3. Use degrees of freedom in q to optimize some criterion or reach some performance

Affine parametrization

Main advantage of YK parametrization:
it is **affine** in the parameter **q**

Take our 4 stable transfer functions

$$\begin{aligned} z_1 &= \frac{bx}{ax + by} u_1 & z_1 &= \frac{-by}{ax + by} u_2 \\ z_2 &= \frac{-by}{ax + by} u_1 & z_2 &= \frac{-ay}{ax + by} u_2 \end{aligned}$$

Recall that

$$\begin{aligned} ax + by &= 1 \\ x &= x_0 + bq \\ y &= y_0 - aq \end{aligned}$$

so that our transfers can be written as

$$\begin{aligned} z_1 &= b(x_0 + bq)u_1 & z_1 &= -b(y_0 - aq)u_2 \\ z_2 &= -b(y_0 - aq)u_1 & z_2 &= -a(x_0 + bq)u_2 \end{aligned}$$

All of them depend **affinely** in the parameter

Convex constraints on transfer functions =
convex constraints on parameter **q**

Parametrization of all stabilizing controllers

Basic example

Integrator plant

$$\frac{b}{a} = \frac{1}{s} = \frac{\left(\frac{1}{s+1}\right)}{\left(\frac{s}{s+1}\right)}$$

One stabilizing controller

$$\frac{y_0}{x_0} = 1$$

particular solution to Bézout equation

$$\frac{s}{s+1}x + \frac{1}{s+1}y = 1$$

All stabilizing controllers

$$\frac{y}{x} = \frac{1 - \frac{s}{s+1}q}{1 + \frac{1}{s+1}q}$$

all solutions to Bézout equation

Another stabilizing controller

$$\frac{y_1}{x_1} = \frac{1 - \frac{s}{s+1} \frac{s+1}{s+2}}{1 + \frac{1}{s+1} \frac{s+1}{s+2}} = \frac{2}{s+3}$$

can be obtained with the choice

$$q = \frac{s+1}{s+2}$$

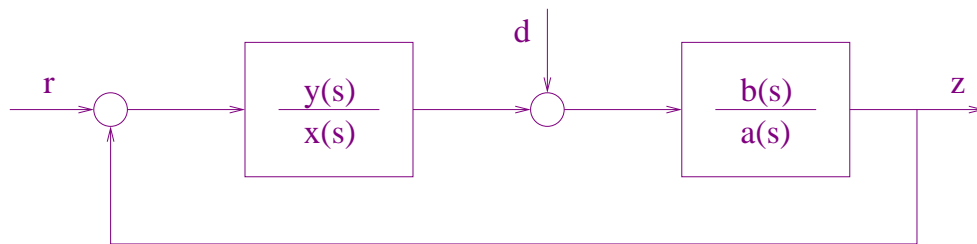
Parametrization of all stabilizing controllers

More interesting example

Consider the plant

$$\frac{b}{a} = \frac{1}{(s-1)(s-2)}$$

in the unity-feedback configuration



We seek a controller y/x ensuring

Spec 1: internal stability

Spec 2: $z \rightarrow 1$ when r unit step and $d = 0$

Spec 3: $z \rightarrow 0$ when d sinusoid 10 rad/s and $r = 0$

All controllers satisfying spec 1 are given by $y/x = (y_0 - aq)/(x_0 + bq)$ for q proper stable rational function

Transfer function from r to $z = b(y - aq)$ hence spec 2 holds iff

$$b(0)(y(0) - a(0)q(0)) = 1$$

Transfer function from d to $z = b(x + bq)$ hence spec 3 holds iff

$$b(j10)(x(j10) + b(j10)q(j10)) = 0$$

Design problem = algebraic problem of finding q such that the above equations hold

Finding the Youla-Kučera parameter

Plant factorization

$$\frac{b}{a} = \frac{\left(\frac{1}{(s+1)^2}\right)}{\left(\frac{(s-1)(s-2)}{(s+1)^2}\right)}$$

Bézout equation $ax + by = 1$ yields a stabilizing controller

$$\frac{y_0}{x_0} = \frac{\left(\frac{19s-11}{s+1}\right)}{\left(\frac{s+6}{s+1}\right)}$$

Algebraic constraints on Youla-Kučera parameter

$$\begin{aligned} q(0) &= 6 \\ \operatorname{Re} q(10j) &= -94 \\ \operatorname{Im} q(10j) &= 70 \end{aligned}$$

Three constraints, so three coefficients to be found

$$q(s) = q_0 + q_1 \frac{1}{s+1} + q_2 \frac{1}{(s+1)^2}$$

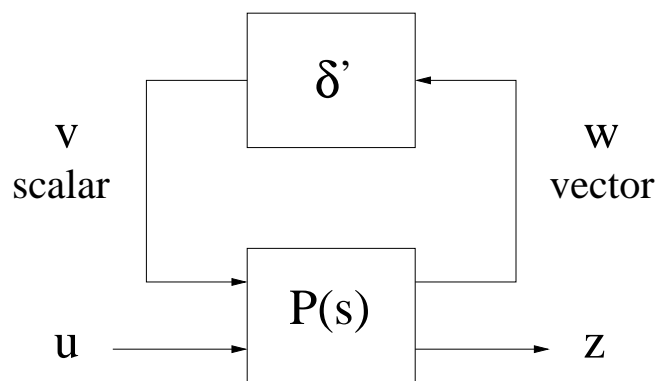
Report the solution $q_0 = -79$, $q_1 = -723$ and $q_2 = 808$ into the controller parametrization

$$\frac{y}{x} = \frac{y_0 - aq}{x_0 + bq} = \frac{-60s^4 - 598s^3 + 2515s^2 - 1794s + 1}{s(s^2 + 100)(s + 9)}$$

Rank-one uncertainty

Using the Youla-Kučera parametrization of all stabilizing controllers, we will now describe a **convex** parametrization of all **robust** controllers obtained by Rantzer and Megretski in 1994

It is valid for a specific (yet useful) class of systems with **rank-one LFT uncertainty**



where uncertainty vector δ satisfies

$$\|\delta\| \leq 1$$

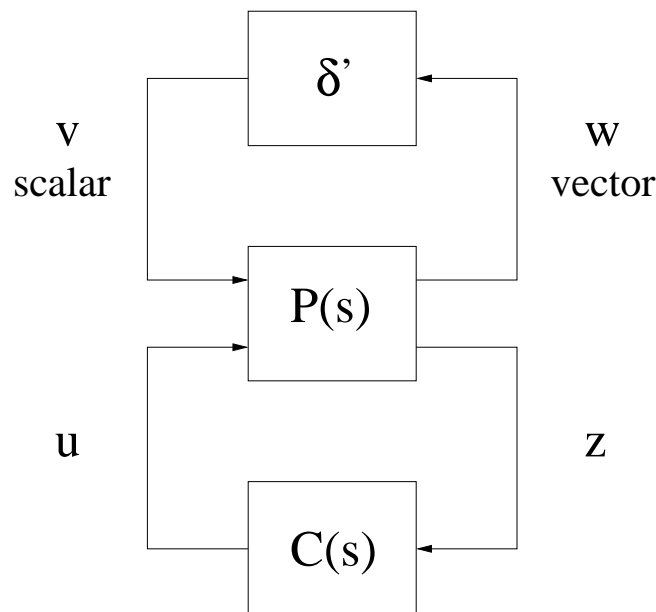
LFT = linear fractional transformation

rank-one = norm-bounded or unstructured

$\|\cdot\| = l_1, l_2$ (Euclidean) or l_∞ norm

Controller parametrization

Now consider the closed-loop system



From the Youla-Kučera parametrization, the transfer function from v to w is given by

$$t_1 + t_2 q$$

where t_1 and t_2 are given elements in R and q is a free parameter in R

Closed-loop system is **robustly stable** iff

$$\frac{1}{1 + \delta'(t_1 + t_2 q)}$$

belongs to R for all admissible uncertainty $\|\delta\| \leq 1$

In this form, **non-convex** condition in q

Convex parametrization

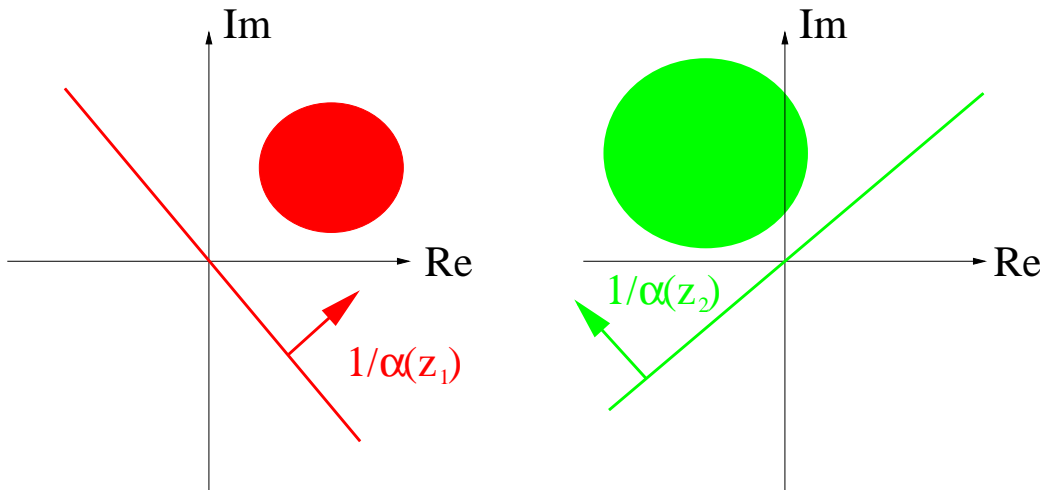
Stability is lost **along stability boundary** D (imaginary axis or unit circle) so the condition

$$\frac{1}{1 + \delta'(t_1 + t_2 q)} \in R \text{ for all } \|\delta\| \leq 1$$

is equivalent to the **zero exclusion condition** on the value set of the rational denominator

$$1 + \delta'(t_1 + t_2 q)(z) \neq 0 \text{ for all } z \in D \text{ for all } \|\delta\| \leq 1$$

For a **fixed frequency** z the above value set V is **convex** hence there exists a **hyperplane** separating V from the origin



Equivalently, for all **fixed** z in D there exists a complex number $\alpha(z)$ in R such that

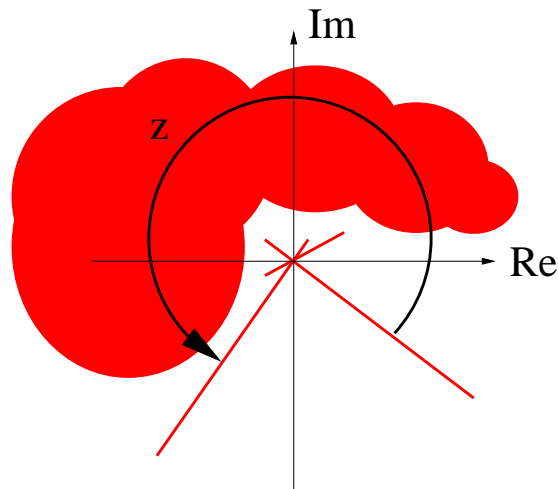
$$\text{Re}(\alpha(z)[1 + \delta'(t_1 + t_2 q)(z)]) > 0 \text{ for all } \|\delta\| \leq 1$$

$\alpha(z)$ is **orthogonal** to the separating hyperplane

Convex parametrization

Only the argument of $\alpha(z)$ matters, so by proper choice of $\|\alpha(z)\|$ (technical part of the proof) it can be shown that this is equivalent to the existence of a $\alpha(z)$ in R valid over the whole frequency range

$$\operatorname{Re}(\alpha(z)[1 + \delta'(t_1 + t_2 q)(z)]) > 0 \text{ for all } z \in D \text{ and } \|\delta\| \leq 1$$



Coprime factorization over R of parameter

$$q(z) = \frac{\beta(z)}{\alpha(z)}$$

yields the equivalent condition

$$\operatorname{Re} \alpha(z) + \delta' \operatorname{Re} (t_1(z)\alpha(z) + t_2(z)\beta(z)) > 0$$

valid for all $z \in D$ and $\|\delta\| \leq 1$

Convex parametrization

The last step is in using the definition of the **dual norm**

$$\|x\|_{\text{dual}} = \max\{x'y : \|y\| \leq 1\}$$

so that the condition

$$\operatorname{Re} \alpha(z) + \delta' \operatorname{Re} (t_1(z)\alpha(z) + t_2(z)\beta(z)) > 0$$

for all $z \in D$ and $\|\delta\| \leq 1$ is equivalent to

$$\|\operatorname{Re} (t_1(z)\alpha(z) + t_2(z)\beta(z))\|_{\text{dual}} < \operatorname{Re} \alpha(z)$$

for all $z \in D$ (dependence on δ is removed)

The above condition is **convex** in design parameters $\alpha(z)$ and $\beta(z)$, so we have proved the main result

The whole set of **robustly stabilizing controllers** for rank-one LFT uncertainties is described by a convex frequency-dependent inequality

We can solve the **frequency dependent** condition with the help of optimization over linear matrix inequalities (**LMIs**), see the last part of the course

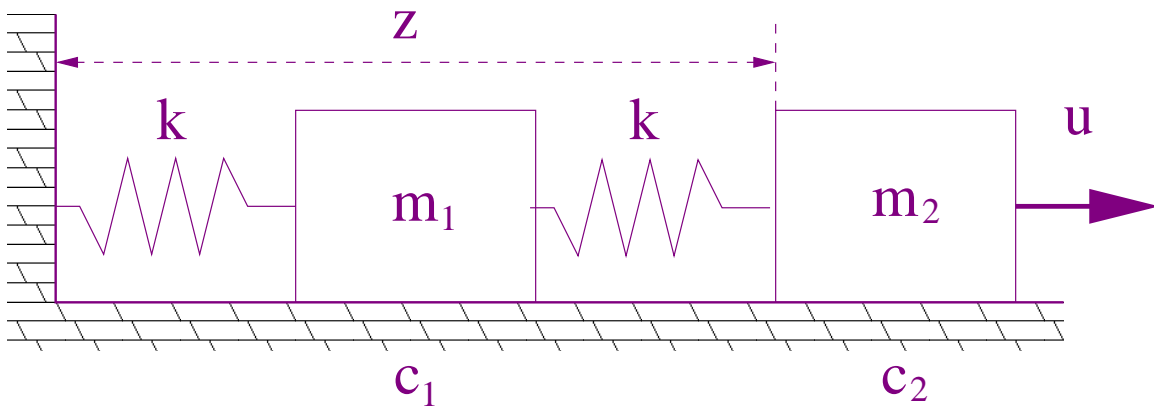
Main strength: the condition is **convex**

Main drawback: the condition is **infinite-dimensional** since the unknowns live in R , and design may result in a controller of **arbitrarily high degree**

Mass-spring system

Example

Consider the mass-spring system



where m_i is the mass, c_i the damping coefficient, k the spring constant, output z the position of the second mass and input u the control force

$$\begin{aligned} m_1 &= 2.25\text{kg}, c_1 = 3.25\text{Ns/m}, k = 423\text{N/m} \\ m_2 &= 2.07\text{kg}, c_2 = 8.18\text{Ns/m} \end{aligned}$$

l_1 norm-bounded uncertainty

$$m_2 = \hat{m}_2 + \delta_1, c_2 = \hat{c}_2 + \delta_2, \|\delta\|_1 \leq 2$$

Open loop transfer function

$$\begin{aligned} \frac{b(s)}{a(s)} &= \frac{k}{(m_1 s^2 + c_1 s + k)(m_2 s^2 + c_2 s + k) - k^2} \\ &= \frac{k}{g_1(s)g_2(s) - k^2} \end{aligned}$$

Mass-spring system

YK controller parametrization

$$\frac{y(s)}{x(s)} = \frac{-k - \frac{g_1 g_2 - k^2}{g_1 g_2} q}{1 - \frac{k}{g_1 g_2} q}$$

Affine closed-loop parametrization

$$t_1(s) + t_2(s)q(s) = \begin{bmatrix} \frac{s^2}{g_2(s)} \\ \frac{s}{g_2(s)} \end{bmatrix} + \begin{bmatrix} -\frac{ks^2}{g_1(s)g_2^2(s)} \\ -\frac{ks}{g_1(s)g_2^2(s)} \end{bmatrix} q(s)$$

Convex parametrization of robustly stabilizing controllers in real ω

$$\left\| \operatorname{Re} \begin{bmatrix} \frac{-\omega^2}{g_2(j\omega)} \\ \frac{j\omega}{g_2(j\omega)} \end{bmatrix} \left(\alpha(j\omega) - \frac{k}{g_1(j\omega)g_2(j\omega)} \beta(j\omega) \right) \right\|_\infty < \operatorname{Re} \alpha(j\omega)$$

Applying a primal-dual LP algorithm, Ghulchak finds the YK parameter

$$q = \frac{-1500s^5 - 2290s^4 + 2599s^3 + 8838s^2 + 8379s + 2933}{s^5 + 2.025s^4 + 1.294s^3 - 1.211s^2 - 2.051s - 1.06}$$

see www.control.lth.se/~ghulchak

Other design methods

Besides Rantzer and Megretski's approach there are several other **robust design methods** based on polynomials or rational fractions and the YK parameter

- mixed sensitivity problem [Kwakernaak]
 H_∞ optimization, spectral factorization
www.math.utwente.nl/~twhuib
- loop shaping [Doyle, Francis et al]
model matching, Nevanlinna-Pick interpolation
www.control.utoronto.ca/people/profs/francis/dft.html
- convex optimization [Boyd]
linear and quadratic programming, LMIs
www.stanford.edu/~boyd

A **finite-dimensional approximation** of the **infinite-dimensional** but **convex** set R of proper stable rational functions must be used when implementing algorithms (convex programming, duality, lower and upper bounds, truncation)