

GRADUATE COURSE ON  
POLYNOMIAL METHODS FOR  
ROBUST CONTROL  
PART III.1

**ROBUST DESIGN:  
POLE PLACEMENT**

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Dancer seated on a pink divan  
Henri de Toulouse-Lautrec (1864-1901)

October-November 2001

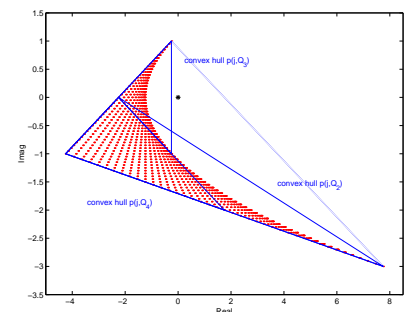
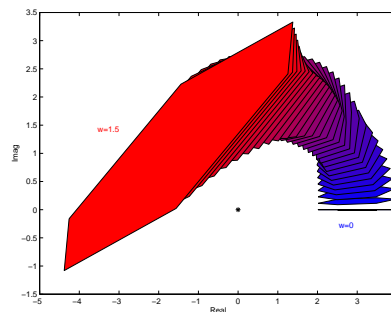
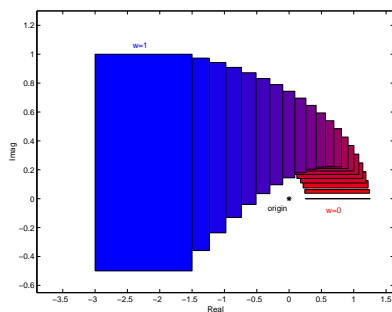
## Summary

In the first part of the course we studied **robust stability analysis**

We saw that checking robust stability can be **easy** (polynomial-time algorithms) or more **difficult** (NP-hard problem), depending namely on the **uncertainty description**

We focused on **polytopic uncertainty**:

- **Interval scalar polynomials**  
Kharitonov's theorem (ct only)
- **Polytope of scalar polynomials**  
(affine polynomial families)  
Edge theorem
- **Interval matrix polynomials**  
(multiaffine polynomial families)  
Mapping theorem
- **Polytopes of matrix polynomials**  
(polynomial polynomial families)



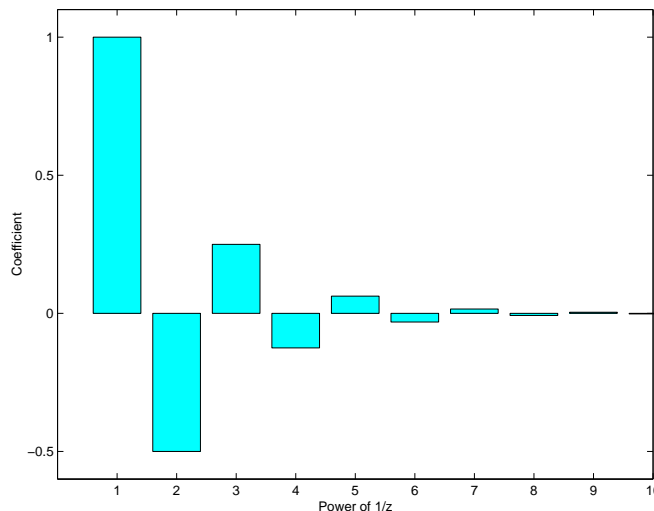
## Robust design

Only in **very special cases** robust analysis techniques can be extended to design  
(for example the 32 plant theorem)

In this third part of the course we will describe robust design algorithms based either on

- **polynomial** pole placement equations, or
- **rational** Youla-Kučera parametrization of stabilizing controllers

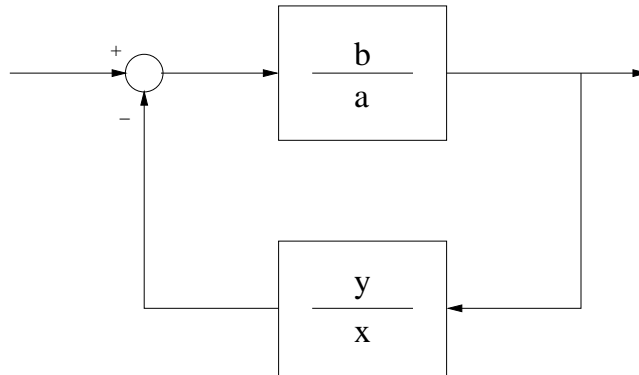
We will see that robust design is a difficult problem when looking for **low-order controllers**



Coefficients of  $1/(1 + 0.5z^{-1}) = r_0 + r_1z^{-1} + r_2z^{-2} + \dots$

## Nominal Pole placement

We consider the SISO feedback system



Closed-loop transfer function

$$\frac{bx}{ax + by}$$

In the absence of hidden modes ( $a$  and  $b$  coprime polynomials), **pole placement** amounts to finding polynomials  $x$  and  $y$  solving the **Diophantine equation** (from Diophantus of Alexandria 200-284)

$$ax + by = c$$

where  $c$  is a given closed-loop characteristic polynomial capturing the **desired system poles**

## Polynomial parametrization of controllers

Let  $x', y'$  be the least degree solution w.r.t  $y'$  of the Diophantine design equation

$$ax + by = c$$

All controllers achieving pole placement are given by

$$\frac{y}{x} = \frac{y' + at}{x' - bt}$$

where  $t$  is a free polynomial **parameter**

### Example

**Water tank** with inflow-to-level transfer function

$$\frac{b}{a} = \frac{1}{s + 1}$$

To counteract level disturbances, a PI controller is sought that places closed-loop poles at  $-6$  and  $-10$ , i.e. we have to solve the Diophantine equation

$$(s + 1)x + y = c = (s + 6)(s + 10)$$

Controllers

$$\frac{y}{x} = \frac{45 + (s + 1)t}{s + 15 - t}$$

PI controller for  $t = 15$

$$\frac{y_0}{x_0} = 15 + \frac{60}{s}$$

## Pole placement

### Example

Consider the rotary hydraulic test rig with identified sampled transfer function

$$\frac{b}{a} = \frac{z^{-3}(-0.0036 + 0.1718z^{-1} + 0.3029z^{-2} - 0.0438z^{-3} - 0.0775z^{-4})}{1 - 2.8805z^{-1} + 3.7827z^{-2} - 2.8269z^{-3} + 1.1785z^{-4} - 0.2116z^{-5}}$$

in the backward shift operator  $z^{-1}$



The desired closed-loop characteristic polynomial is

$$c = (1 - 0.3z^{-1})(1 - 0.4z^{-1}) = 1 - 0.7z^{-1} + 0.12z^{-2}$$

Solving the **Diophantine equation**  $ax + by = c$  yields the **unstable** controller

$$\frac{y'}{x'} = \frac{2.9023 - 6.7682z^{-1} + 7.4670z^{-2} - 4.3287z^{-3} + 1.0169z^{-4}}{1 + 2.1805z^{-1} + 2.6182z^{-2} + \dots - 0.3725z^{-6}}$$

With the choice

$$t = 1.2222 - 0.1952z^{-1} - 0.1310z^{-2} + 0.5663z^{-3} + 0.8805z^{-4} + 0.5677z^{-5}$$

as a free polynomial parameter we obtain a **stable** controller of higher order

$$\frac{y''}{x''} = \frac{1.6801 - 3.0525z^{-1} + 2.4125z^{-2} + \dots + 0.1201z^{-10}}{1 + 2.1085z^{-1} + 2.6182z^{-2} + \dots - 0.0440z^{-12}}$$

## Pole placement: numerical aspects

The polynomial Diophantine equation

$$ax + by = c$$

is **linear** in unknowns  $x$  and  $y$ , and denoting

$$\begin{aligned} a(s) &= a_0 + a_1s + \dots + a_{d_a}s^{d_a} \\ x(s) &= x_0 + x_1s + \dots + x_{d_b}s^{d_b} \\ &\text{etc..} \end{aligned}$$

we can **identify powers** of the indeterminate  $s$  to build a **linear system of equations**

$$\left[ \begin{array}{ccc|ccc} a_0 & & & b_0 & & \\ & \ddots & & b_1 & \ddots & \\ & & a_0 & \vdots & & b_0 \\ a_{d_a} & & a_1 & b_{d_b} & & b_1 \\ & \ddots & \vdots & & \ddots & \vdots \\ & & a_{d_a} & & & b_{d_b} \end{array} \right] \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ \frac{x_{d_x}}{y_0} \\ y_1 \\ \vdots \\ y_{d_y} \end{bmatrix} = \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_{d_c} \end{bmatrix}$$

The above matrix is called **Sylvester matrix**, it has a special Toeplitz banded structure that can be exploited when solving the equation



James J Sylvester  
(1814 London - 1897 London)



Otto Toeplitz  
(1881 Breslau - 1940 Jerusalem)

## Pole placement for MIMO systems

Pole placement can be performed similarly for a plant left MFD

$$A^{-1}(s)B(s)$$

with a controller right MFD

$$Y(s)X^{-1}(s)$$

The Diophantine equation to be solved is now over [polynomial matrices](#)

$$A(s)X(s) + B(s)Y(s) = C(s)$$

and right hand-side matrix  $C(s)$  captures information on invariant polynomials and eigenstructure

For example  $C(s)$  may contain  $H_2$  or  $H_\infty$  optimal dynamics (obtained with spectral factorization)



## Robust pole placement

Now assume that the plant transfer function

$$\frac{b(q)}{a(q)}$$

contains some uncertain parameter  $q$

The problem of robust pole placement will then consist in finding a controller

$$\frac{y}{x}$$

such that the uncertain closed-loop characteristic polynomial

$$a(q)x + b(q)y = c(q)$$

is robustly stable

How can we find  $x, y$  to ensure robust stability of  $c(q)$  for all admissible uncertainty  $q$  ?

Coefficients of  $c$  are linear in  $x$  and  $y$ , but are stability conditions also linear in  $c$  ?

## Stability criteria

Given a continuous-time polynomial

$$p(s) = p_0 + p_1s + \cdots + p_{n-1}s^{n-1} + p_ns^n$$

with  $p_n > 0$  we define its  $n \times n$  **Hurwitz matrix**

$$H(p) = \begin{bmatrix} p_{n-1} & p_{n-3} & & 0 & 0 \\ p_n & p_{n-2} & & \vdots & \vdots \\ 0 & p_{n-1} & \cdots & 0 & 0 \\ 0 & p_n & & p_0 & 0 \\ \vdots & \vdots & & p_1 & 0 \\ 0 & 0 & & p_2 & p_0 \end{bmatrix}$$

**Hurwitz stability criterion:** Polynomial  $p(s)$  is stable iff all principal minors of  $H(p)$  are  $> 0$

### Example

For  $n = 4$  the stability conditions are

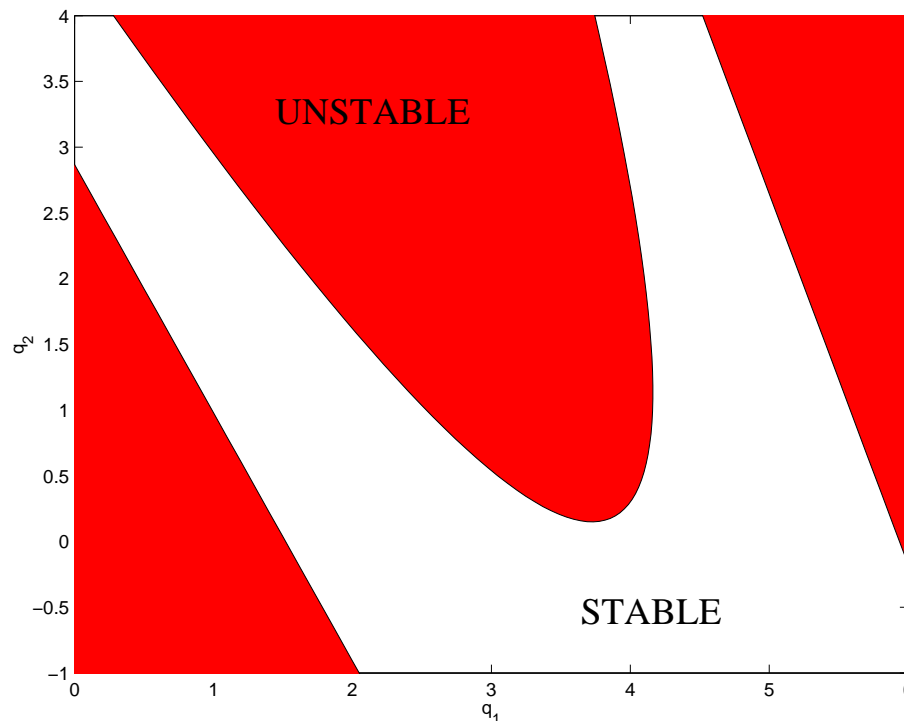
$$\begin{aligned} p_3 &> 0 \\ p_2p_3 - p_1p_4 &> 0 \\ p_1p_2p_3 - p_0p_3^2 - p_1^2p_4 &> 0 \\ p_0p_1p_2p_3 - p_0^2p_3^2 - p_0p_1^2p_4 &> 0 \end{aligned}$$

Highly non-linear inequalities

Similar criteria for discrete-time polynomials (Jury matrix) or any other **stability region** (sector, parabola etc..)

## Non-convexity of stability domain

Main problem: the stability domain in the space of polynomial coefficients  $p_i$  is **non-convex** in general



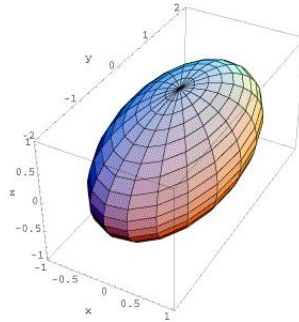
Discrete-time stability domain in  $(q_1, q_2)$  plane for polynomial  $p(z, q) = (-0.825 + 0.225q_1 + 0.1q_2) + (0.895 + 0.025q_1 + 0.09q_2)z + (-2.475 + 0.675q_1 + 0.3q_2)z^2 + z^3$

How can we **overcome** the non-convexity of the stability conditions in the coefficient space ?

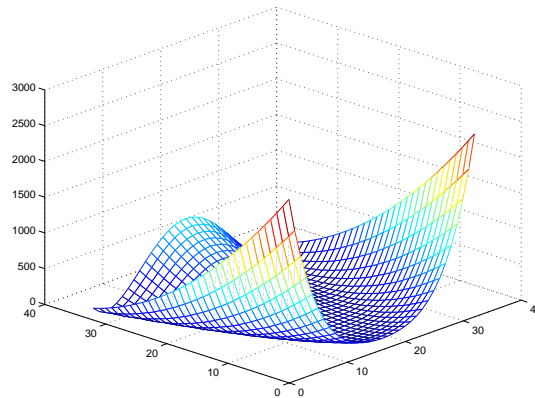
## Handling non-convexity

Basically, we can pursue two approaches:

- we can **approximate** the non-convex stability domain with a convex domain (segment, polytope, sphere, ellipsoid, LMI)



- we can address the non-convexity with the help of **non-convex optimization** (global or local optimization)



## Approximation of the stability domain

From the tools of **robust stability analysis**, we can build around a nominally stable polynomial

- stability **segments** (eigenvalue criterion)
- stability **boxes** (Kharitonov's theorem)
- stability **polytopes** (Edge theorem)

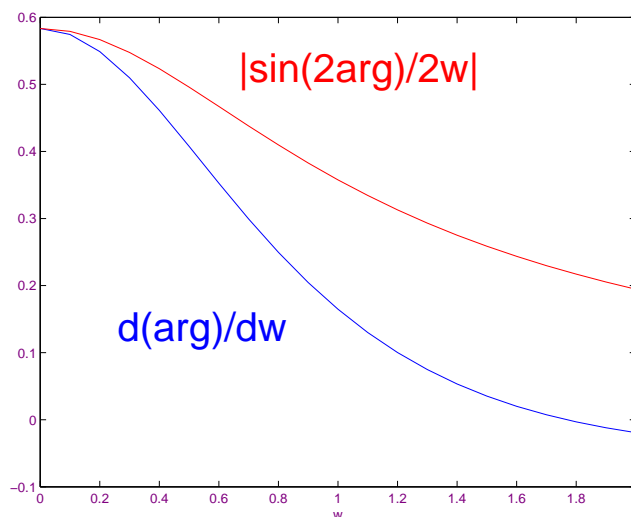
There exists other results, such as **Rantzer's growth condition**: a polynomial  $g(s)$  is a convex direction iff

$$\frac{d}{d\omega} \arg g(j\omega) \leq \left| \frac{\sin 2 \arg g(j\omega)}{2\omega} \right|, \quad \omega > 0$$

It means that given any stable  $f(s)$  such that  $f(s) + g(s)$  is stable then the whole segment  $[f(s), g(s)]$  is stable

### Example

$g(s) = 24 + 14s - 13s^2 - 2s^3 + s^4$  is a growth direction



## Stability polytopes

Largest **hyper-rectangle** around a nominally stable polynomial

$$p(s) + r \sum_{i=0}^n [-\varepsilon_i, \varepsilon_i] s^i$$

obtained with the **eigenvalue criterion** applied on the 4 **Kharitonov polynomials**

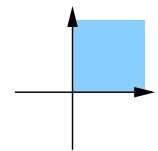
In general, there is no **systematic way** to obtain more general stability polytopes, namely because of computational complexity

(no analytic formula for the volume of a polytope)

Well-known candidates:

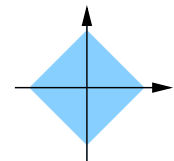
- ct LHP: outer approximation (necessary stab cond)

**positive cone**  $p_i > 0$



- dt unit disk: inner approximation (sufficient stab cond)

**diamond**  $|p_0| + |p_1| + \dots + |p_{n-1}| < 1$



## Stability polytopes

**Necessary** stab cond in dt: convex hull of stability domain is a polytope whose  $n + 1$  vertices are polynomials with roots  $+1$  or  $-1$

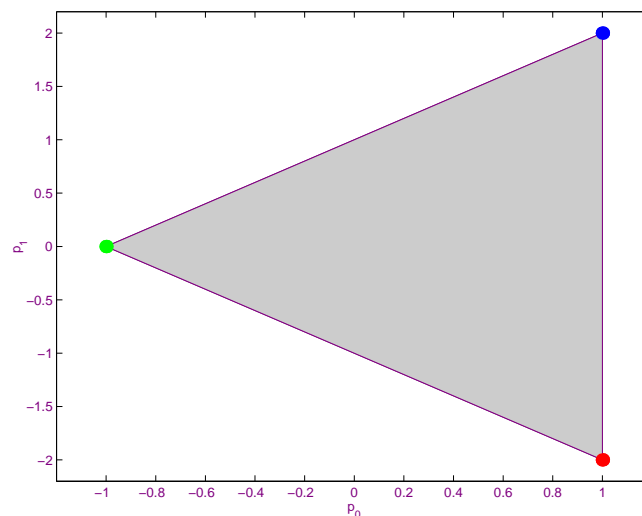
### Example

When  $n = 2$ : triangle with vertices

$$(z + 1)(z + 1) = 1 + 2z + z^2$$

$$(z + 1)(z - 1) = -1 + z^2$$

$$(z - 1)(z - 1) = 1 - 2z + z^2$$



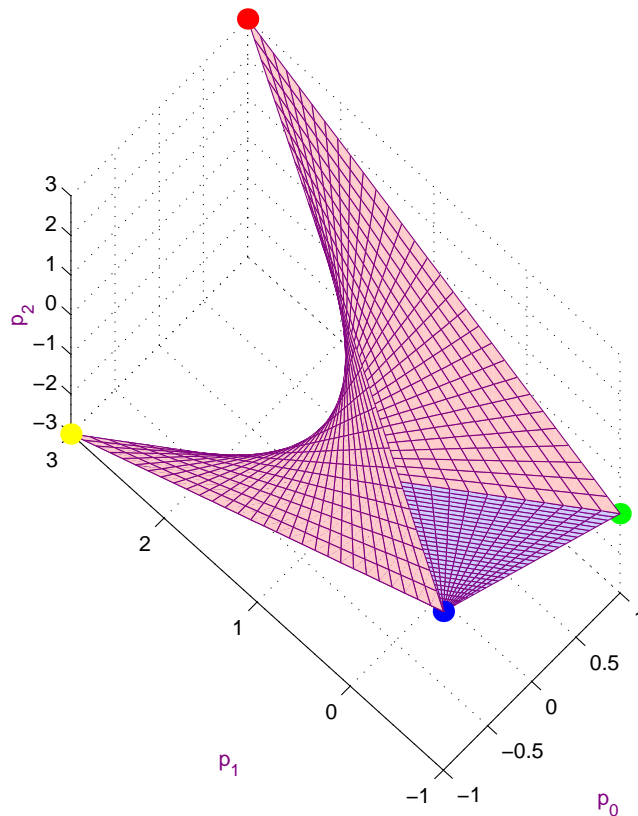
In this simple case it is the exact stability domain

## Stability polytopes

### Example

Third degree dt polynomial: two hyperplanes and a **non-convex** hyperbolic paraboloid with a saddle point at  $p(z) = p_0 + p_1z + p_2z^2 + z^3 = z(1 + z^2)$

$$\begin{aligned} (z + 1)(z + 1)(z + 1) &= 1 + 3z + 3z^2 + z^3 \\ (z + 1)(z + 1)(z - 1) &= -1 - z + z^2 + z^3 \\ (z + 1)(z - 1)(z - 1) &= 1 - z - z^2 + z^3 \\ (z - 1)(z - 1)(z - 1) &= -1 + 3z - 3z^2 + z^3 \end{aligned}$$



The convex hull is made of four hyperplanes



## Stability hyper-spheres

Largest **hyper-sphere** around a nominally stable polynomial

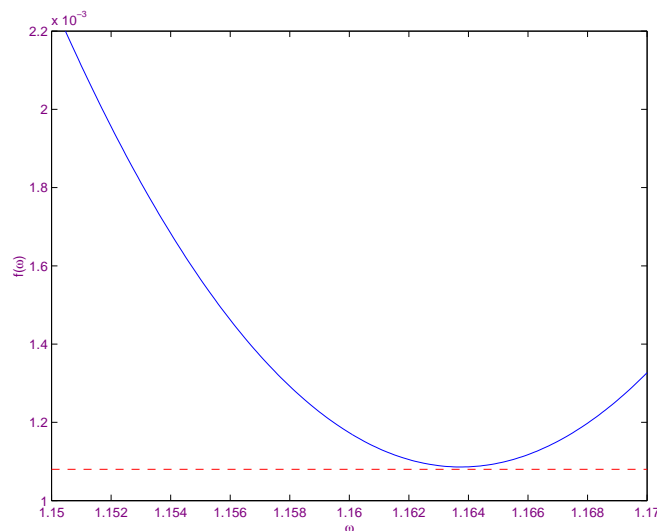
$$p(s) + \sum_{i=0}^n q_i s^i, \|q\| \leq r$$

has radius

$$r_{\max} = \min\{|p_0|, |p_n|, \inf_{\omega>0} \sqrt{\frac{(\operatorname{Re} p(j\omega))^2}{1 + w^4 + w^8 \dots} + \frac{(\operatorname{Im} p(j\omega))^2}{w^2 + w^6 + \dots}}\}$$

### Example

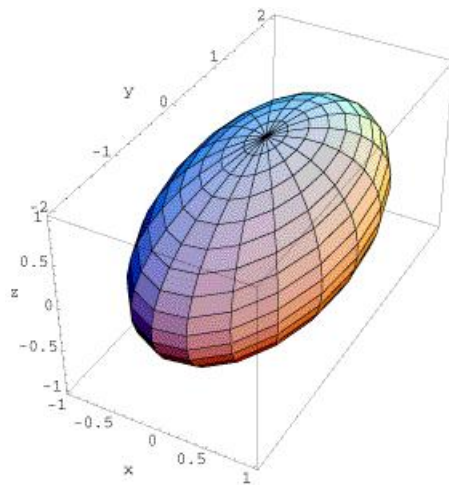
$$(2 + q_0) + (1.4 + q_1)s + (1.5 + q_2)s^2 + (1 + q_3)s^3, \|q\| \leq r$$



$$r_{\max} = \min\{2, 1, \inf_{\omega>0} f(w)\} = 1.08 \cdot 10^{-3}$$

## Stability ellipsoids

A **weighted** and **rotated** hyper-sphere is an **ellipsoid**



Using recently developed algorithms based on **LMI optimization** one can obtain ellipsoidal inner approximation of stability domains

$$E = \{p : (p - \bar{p})^T P (p - \bar{p}) \leq 1\}$$

where

- $p$  coef vector of polynomial  $p(s)$
- $\bar{p}$  center of ellipsoid
- $P$  positive definite matrix

## Stability ellipsoids

### Example

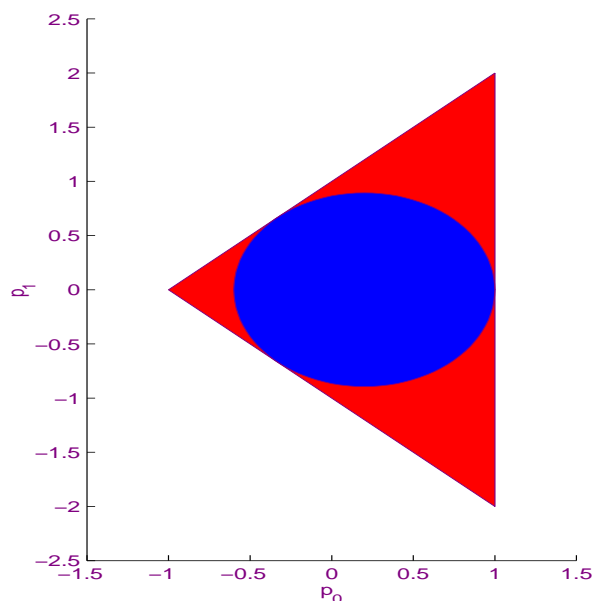
Discrete-time second degree polynomial

$$p(z) = p_0 + p_1z + z^2$$

We solve the LMI problem and we obtain

$$P = \begin{bmatrix} 1.5625 & 0 \\ 0 & 1.2501 \end{bmatrix} \quad \bar{p} = \begin{bmatrix} 0.2000 \\ 0 \end{bmatrix}$$

which describes an ellipse  $E$  inscribed in the exact triangular stability domain  $S$



## Stability ellipsoids

### Example

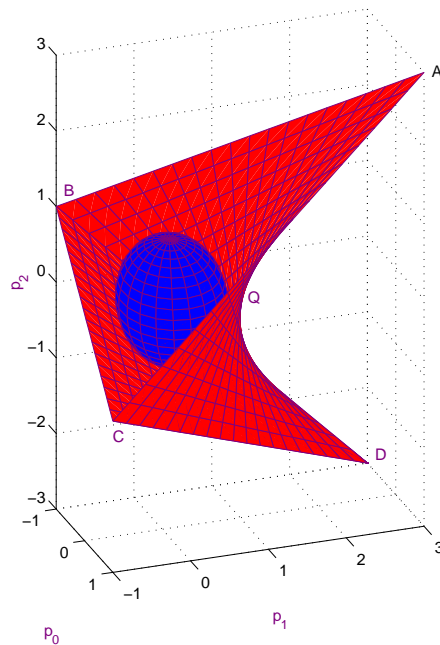
Discrete-time third degree polynomial

$$p(z) = p_0 + p_1z + p_2z^2 + z^3$$

We solve the LMI problem and we obtain

$$P = \begin{bmatrix} 2.3378 & 0 & 0.5397 \\ 0 & 2.1368 & 0 \\ 0.5397 & 0 & 1.7552 \end{bmatrix} \quad \bar{x} = \begin{bmatrix} 0 \\ 0.1235 \\ 0 \end{bmatrix}$$

which describes a **convex** ellipse  $E$  inscribed in the exact stability domain  $S$  delimited by the **non-convex** hyperbolic paraboloid



**Very simple** scalar sufficient stability condition

$$2.4166p_0^2 + 2.2088p_1^2 + 1.8143p_2^2 - 0.5458p_1 + 1.1158p_0p_2 \leq 1$$

## Volume of stability ellipsoid

In the discrete-time case, the well-known sufficient stability condition defines a diamond

$$R = \{p : |p_0| + |p_1| + \cdots + |p_{n-1}| < 1\}$$

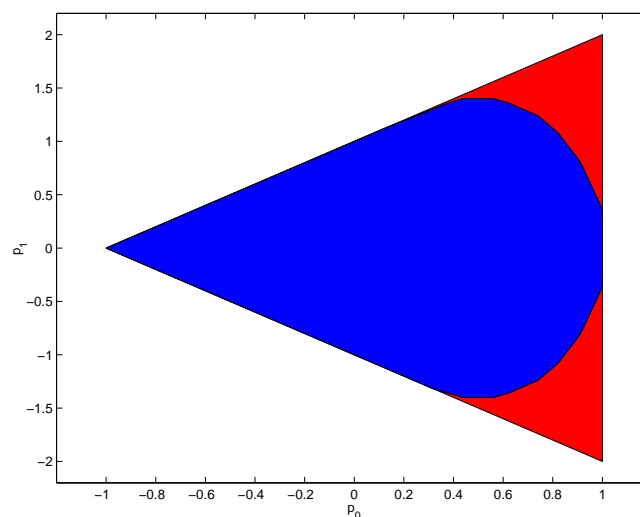
For different values of degree  $n$ , we compared **volumes** of exact stability domain  $S$ , ellipsoid  $E$  and diamond  $D$

	$n = 2$	$n = 3$	$n = 4$	$n = 5$
Stability domain $S$	4.0000	5.3333	7.1111	7.5852
Ellipsoid $E$	2.2479	1.4677	0.7770	0.3171
Diamond $D$	2.0000	1.3333	0.6667	0.2667

$E$  is “larger” than  $D$ , yet very small wrt  $S$

## LMI stability regions

In the fourth and last part of this course, we will propose better **LMI** inner approximations of the stability domain



## Robust pole placement

Once we have a convex approximation of the stability region, we can perform design with

- linear programming (polytopes)
- quadratic programming (spheres, ellipsoids)
- semidefinite programming (LMIs)

Complexity of design algorithm **increases**

Conservatism of control law **decreases**



## Application in robust design

### First example

MIMO plant with right MFD

$$B(s)A^{-1}(s) = \begin{bmatrix} b & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} s+1 & 0 \\ 0 & s+1 \end{bmatrix}^{-1}$$

with uncertainty in parameter

$$b \in [0.5, 1.5]$$

We seek a proper first order controller

$$X^{-1}(s)Y(s) = \begin{bmatrix} s+x_1 & x_2 \\ 0 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} y_1s+y_2 & y_3s+y_4 \\ 0 & y_5 \end{bmatrix}$$

assigning robustly the closed-loop polynomial matrix

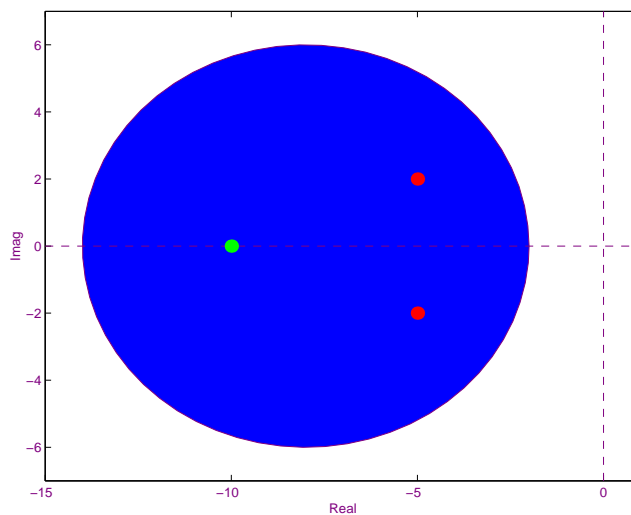
$$C(s) = \begin{bmatrix} s^2 + \alpha s + \beta & \delta(s) \\ 0 & s + \gamma \end{bmatrix}$$

whose coefficients live in the polytope

$$\begin{bmatrix} -14 & 1 & 0 \\ 16 & -2 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} > \begin{bmatrix} -196 \\ 56 \\ -4 \\ 2 \\ -14 \end{bmatrix}$$

These specifications amounts to assigning the poles within the disk

$$|s+8| < 6$$



## Application in robust design

### First example (end)

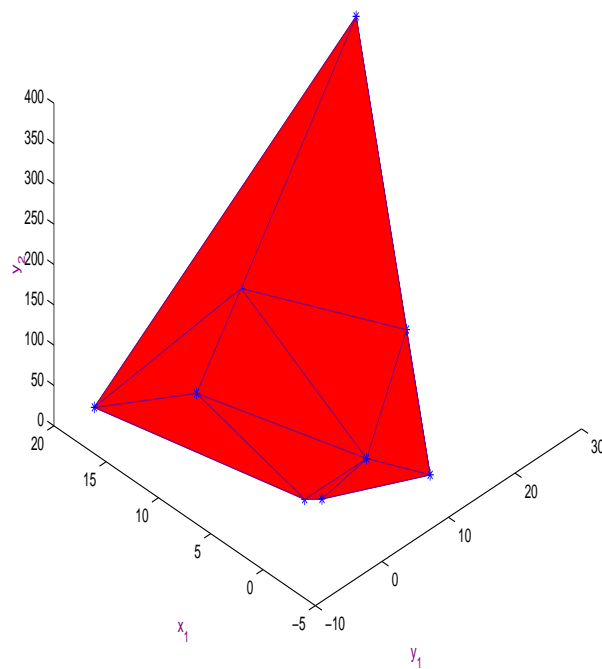
Equating powers of indeterminate  $s$  in the polynomial matrix Diophantine equation

$$X(s)A(s) + Y(s)B(s) = C(s)$$

we obtain the design inequalities

$$\begin{bmatrix} -13 & -7 & 0.5 \\ 14 & 8 & -1 \\ -1 & -1 & 0.5 \\ -13 & -21 & 1.5 \\ 14 & 24 & -3 \\ -1 & -3 & 1.5 \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \\ y_2 \end{bmatrix} > \begin{bmatrix} -182 \\ 40 \\ -2 \\ -182 \\ 40 \\ -2 \end{bmatrix}$$

characterizing all parameters  $x_1$ ,  $y_1$  and  $y_2$  of admissible robust controllers



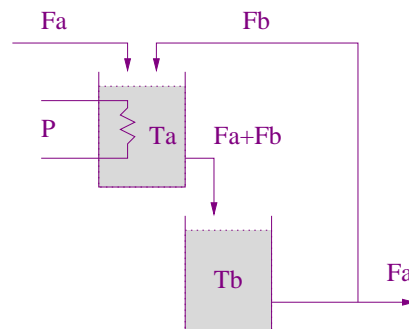
Corresponding polytope with 9 vertices



## Application in robust design

### Second example

We consider two mixing tanks arranged in cascade with recycle stream



The controller must be designed to maintain temp  $T_b$  of the second tank at desired set point by manipulating power  $P$  delivered by heater located in first tank - only available measurement is temp  $T_b$

Nominal plant identification is carried out with least-squares method, resulting in 2nd order dt model

$$\frac{b(z)}{a(z)} = \frac{a_1 + a_2 z}{a_3 + a_4 z + z^2}$$

with nominal plant vector

$$\bar{a} = [ 0.0038 \quad 0.0028 \quad 0.2087 \quad -1.1871 ]$$

Positive definite matrix  $Q$  characterizing the uncertainty ellipsoid  $\{a : (a - \bar{a})' Q (a - \bar{a}) \leq 1\}$  is readily available as a by-product of the identification technique

## Application in robust design

### Second example (end)

We perform robust pole placement in the inner ellipsoidal approximation

$$\{p : (p - \bar{p})'P(p - \bar{p}) \leq 1\}$$

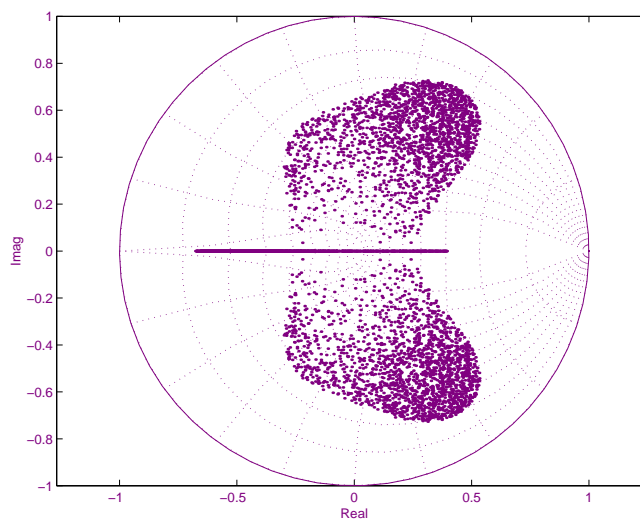
of the 3rd degree non-convex stability region, with

$$P = \begin{bmatrix} 2.3378 & 0 & 0.5397 \\ 0 & 2.1368 & 0 \\ 0.5397 & 0 & 1.7552 \end{bmatrix} \quad \bar{p} = \begin{bmatrix} 0 \\ 0.1235 \\ 0 \end{bmatrix}$$

We solve a convex quadratic optimization problem to obtain a first-order controller

$$\frac{y(z)}{x(z)} = \frac{0.3377 + 166.0z}{0.6212 + z}$$

robustly stabilizing the plant



Robust closed-loop root-locus for random admissible ellipsoidal uncertainty