

LP Infeasibility - The Pursuit of Intelligent Discretization

WORKSHOP ON CONVEX AND REAL-TIME OPTIMIZATION

August 19, 2016

Tobias Leth
tol@es.aau.dk

Department of Electronic Systems
Aalborg University
Denmark



AALBORG UNIVERSITY
DENMARK

Agenda



Introduction

Bernstein Basis

Polynomials in the Bernstein Basis

Lyapunov Functions in the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint Identification

Farkas' Lemma

Open Questions

CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Lyapunov Fcn for Stability Analysis



CortOpt

Tobias Leth

2 Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Classical Lyapunov criteria:

$$V(x) > 0 \wedge V(0) = 0,$$

$$\dot{V}(x) \leq 0.$$

Thus V is positive definite (PD) and $-\dot{V}$ is either PS or positive semi-definite (PSD).

Lyapunov Fcn for Stability Analysis



CortOpt

Tobias Leth

Classical Lyapunov criteria:

$$V(x) > 0 \wedge V(0) = 0,$$
$$\dot{V}(x) \leq 0.$$

Current standard is to relax as

$$V(x) \in \Sigma^2$$
$$-\dot{V}(x) \in \Sigma^2$$

for global analysis, or as

$$V(x) - \sum_i s_i p_i \in \Sigma^2$$
$$-\dot{V}(x) - \sum_i s_i p_i \in \Sigma^2$$

for local analysis using Putinar's Positivstellensatz.

3 Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Another Way



CortOpt

Tobias Leth

4 Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

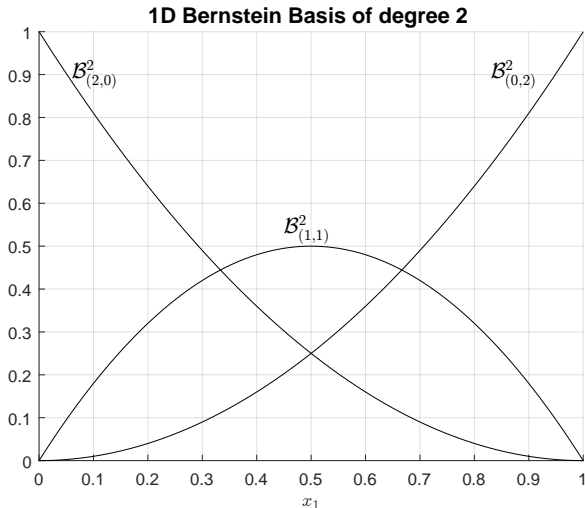
But $\text{PSD} \neq \text{SOS}$. And solving the SOS programs require semi-definite programming (SDP).

We do something else. Our method relies on the Bernstein basis. It always results in local analysis, but is also always results in linear programming (LP).

Bernstein basis Polynomials



Defined on a simplex σ : 1D is a line, 2D is a triangle, 3D and up is called a simplex.



CortOpt

Tobias Leth

Introduction

5 Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Dept. of Electronic Systems
Aalborg University
Denmark

Bernstein basis Polynomials



CortOpt

Tobias Leth

Introduction

6 Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

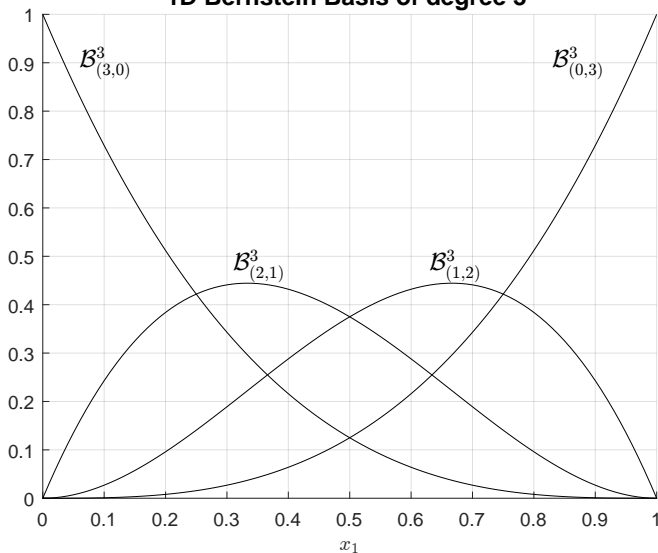
Constraint
Identification

Farkas' Lemma

Open Questions

Dept. of Electronic Systems
Aalborg University
Denmark

1D Bernstein Basis of degree 3



Bernstein basis Polynomials



CortOpt

Tobias Leth

Introduction

7 **Bernstein Basis**

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

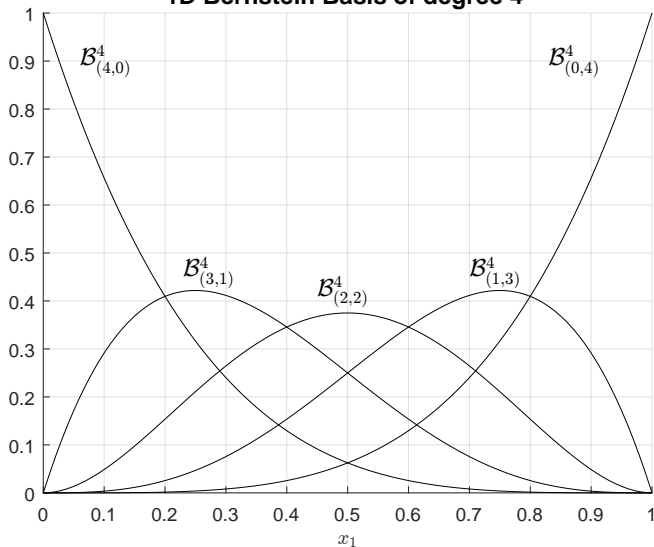
Constraint
Identification

Farkas' Lemma

Open Questions

Dept. of Electronic Systems
Aalborg University
Denmark

1D Bernstein Basis of degree 4



Bernstein basis Polynomials



CortOpt

Tobias Leth

Introduction

8 **Bernstein Basis**

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

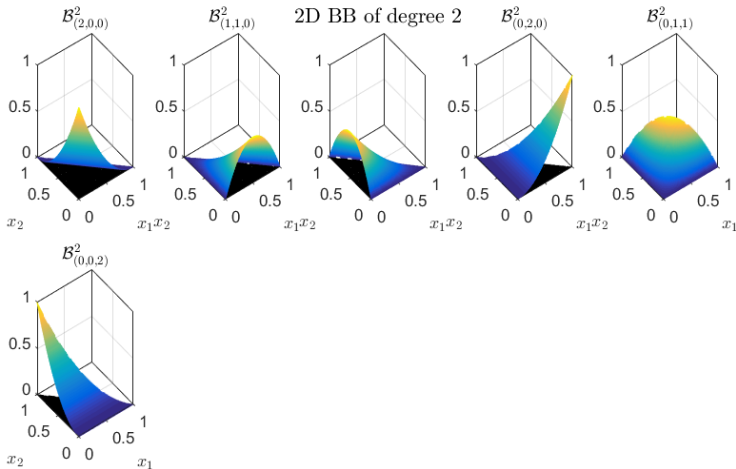
Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Dept. of Electronic Systems
Aalborg University
Denmark



Bernstein basis Polynomials



CortOpt

Tobias Leth

Introduction

9 **Bernstein Basis**

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

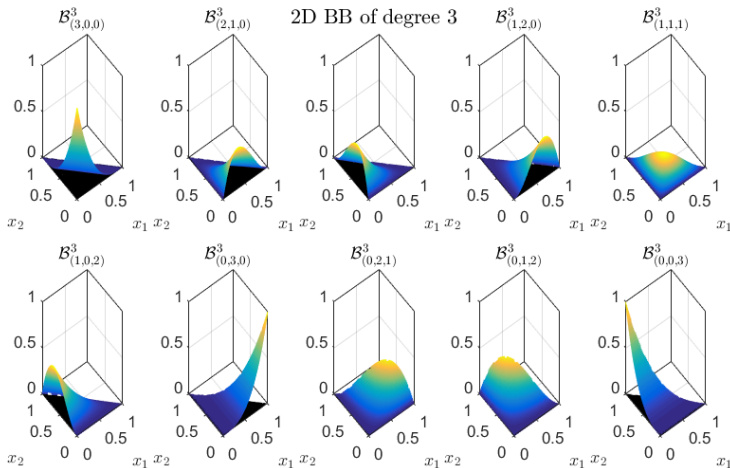
Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions



Bernstein basis Polynomials



CortOpt

Tobias Leth

Introduction

10 **Bernstein Basis**

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

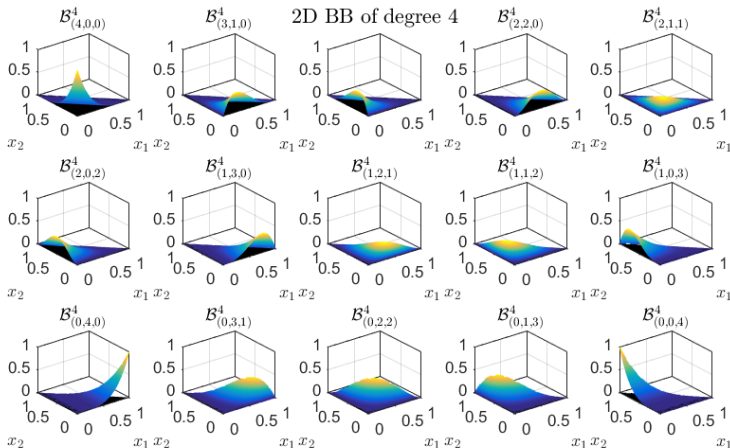
Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions



Bernstein Coefficients



CortOpt

Tobias Leth

Introduction

Bernstein Basis

11 Polynomials in the Bernstein Basis

Lyapunov Functions in the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint Identification

Farkas' Lemma

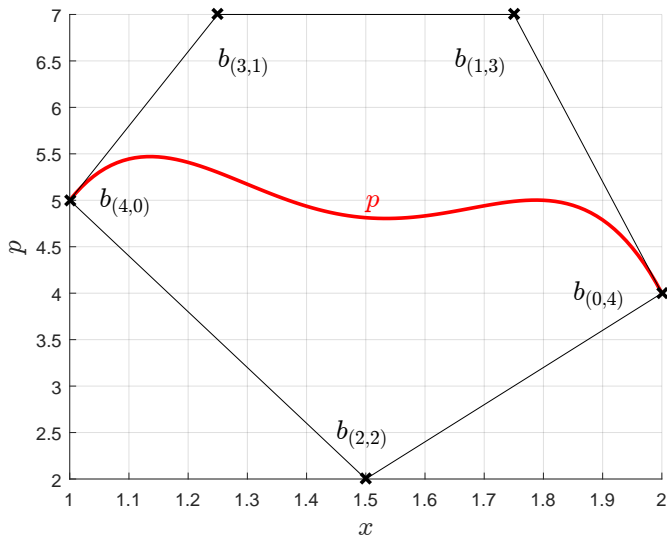
Open Questions

Any polynomial can be described in the Bernstein basis by a linear combination of the Bernstein basis polynomials, \mathcal{B}_α^d .

$$p = \sum_{|\alpha|=d} b_\alpha(p, d, \sigma) \mathcal{B}_\alpha^d$$

The Bernstein coefficients contain valuable information about the polynomial p . The basis polynomials are non-negative on the simplex. Thus, if the coefficients are positive then so is p on the simplex.

Convex Hull & End-point Value Property



CortOpt

Tobias Leth

Introduction

Bernstein Basis

12 Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

Dept. of Electronic Systems
Aalborg University
Denmark

Classical Lyapunov criteria



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

13 Lyapunov Functions in
the Bernstein Basis

Example

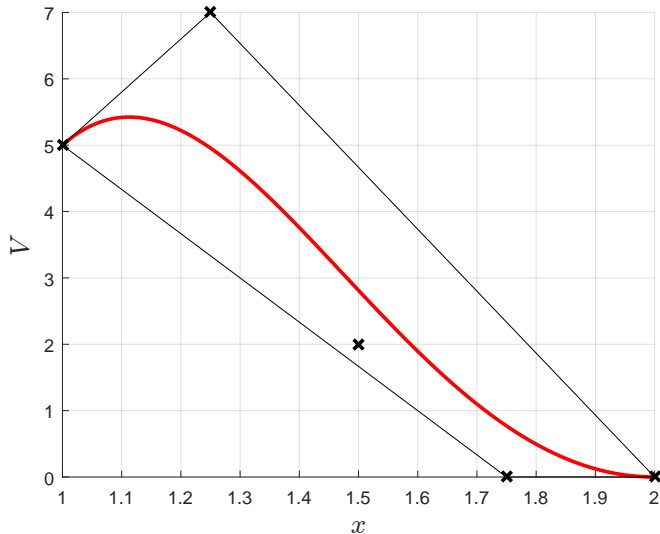
Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions



Now to the Fun of it!



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

14 Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

$$V = \sum_{|\alpha_V|=d_V} CV_{\alpha_V} \mathcal{B}_{\alpha_V}^{d_V} = CV^T \mathcal{B}^{d_V}$$

$$\dot{V} = \sum_{|\hat{\alpha}|=\hat{d}} CL_{\hat{\alpha}} \mathcal{B}_{\hat{\alpha}}^{\hat{d}} = CL^T \mathcal{B}^{\hat{d}}$$

$$CV \geq 0$$

$$CV_{\alpha_V^*} = 0$$

$$CL \leq 0$$

$$CL_{\hat{\alpha}^*} = 0$$

Now to the Fun of it!



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

15 Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

$$V = \sum_{|\alpha_V|=d_V} CV_{\alpha_V} \mathcal{B}_{\alpha_V}^{d_V} = CV^T \mathcal{B}^{d_V}$$

$$\dot{V} = \sum_{|\hat{\alpha}|=\hat{d}} CL_{\hat{\alpha}} \mathcal{B}_{\hat{\alpha}}^{\hat{d}} = CL^T \mathcal{B}^{\hat{d}}$$

$$CV \geq 0$$

$$CV_{\alpha_V^*} = 0$$

$$CL \leq 0$$

$$CL_{\hat{\alpha}^*} = 0$$

$$CL = A CV$$

Linear Feasibility Problem



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

16 Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

$$\begin{aligned} \min_{CV} \quad & 0 \\ \text{s.t.} \quad & I^c \leq A CV \leq 0 \\ & 0 \leq CV \leq u^x \end{aligned}$$

Any CV solving the LP problem is a Lyapunov function for the vector field and certifies the local stability. Infeasibility does not suggest the converse!

The Vector Field



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

17 **Example**

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

$$\dot{x}_1 = -x_1^3 x_2^2 + 2x_1^3 x_2 - x_1^3 + 4x_1^2 x_2^2 - 8x_1^2 x_2 \\ + 4x_1^2 - x_1 x_2^4 + 4x_1 x_2^3 - 4x_1 + 10x_2^2$$

$$\dot{x}_2 = -9x_1^2 x_2 + 10x_1^2 + 2x_1 x_2^3 - 8x_1 x_2^2 - 4x_1 - x_2^3 + 4x_2^2 - 4x_2$$

Two variables, degree 5, origin is stable and it is the only equilibrium.

Investigated on the box $B = [\pm 1]^2$ with a Lyapunov function of degree $d_V = 2$.

The Solution



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

18 **Example**

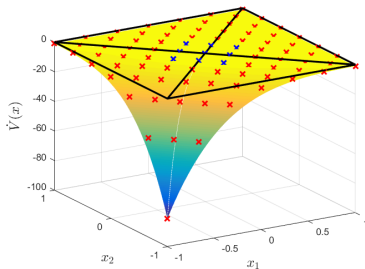
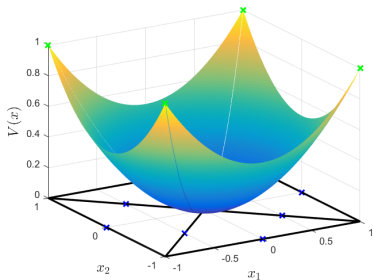
Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions





CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

19 **Infeasibility**

Example Continued

Constraint
Identification

Farkas' Lemma

Open Questions

But if the box is expanded to $B = [\pm 1.75]^2$ the resulting LP is unsolvable. Then what?

Bernstein's Theorem

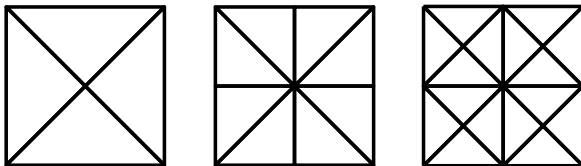


CortOpt

Tobias Leth

Bernstein's Theorem: If a polynomial p of degree d is positive on a simplex σ , then there exists a sub-division of σ into a collection $K = \{\sigma^1, \dots, \sigma^m\}$ of finitely many simplices such that

$$b(p, d, \sigma^i) > 0 \forall i \in \{1, \dots, m\}.$$



This motivates the use of sub-division in order to obtain a solvable LP.

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

20 Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

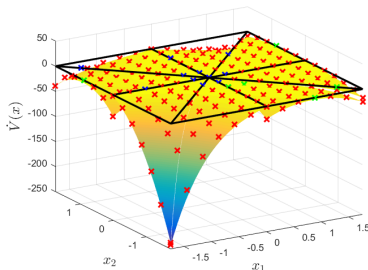
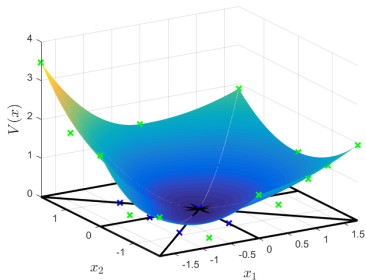
Open Questions

The Solution



CortOpt
Tobias Leth

Sub-dividing into 8 simplices renders the LP solvable.



But the number of design variables and constraints grow rapidly with increase in the number of simplices.

- Introduction
- Bernstein Basis
- Polynomials in the Bernstein Basis
- Lyapunov Functions in the Bernstein Basis
- Example
- Inf feasibility
- 21 Example Continued
- Constraint Identification
- Farkas' Lemma
- Open Questions

Adding Slack Variables



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

22 **Constraint
Identification**

Farkas' Lemma

Open Questions

By adding one non-negative slack variable on each simplex, and minimizing their sum, the problematic parts of the domain can be identified

$$\begin{aligned} \min_{CV, s_i} \quad & \sum_i^m s_i \\ \text{s.t.} \quad & l^c \leq A CV - Bs \leq 0 \\ & 0 \leq CV \leq u^x \\ & 0 \leq s_i \end{aligned}$$

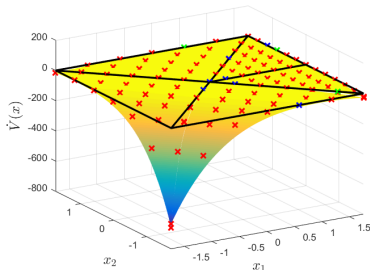
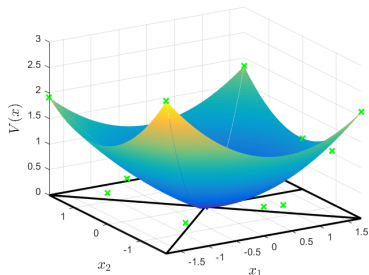
This singles out some simplices for further sub-division, and leaves the rest as they are.

The Solution



CortOpt
Tobias Leth

And it works.



Saves on the number of design variables and constraints, yet obtains a solvable LP.

Introduction
Bernstein Basis
Polynomials in the Bernstein Basis
Lyapunov Functions in the Bernstein Basis
Example
Infeasibility
Example Continued
23 Constraint Identification
Farkas' Lemma
Open Questions



Everything so far is covered in the paper:

T. Leth, C. Sloth, and R. Wisniewski, "Lyapunov Function Synthesis - Algorithm and Software," *MSC, Buenos Aries, Argentina*, 2016.

CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

24 **Constraint
Identification**

Farkas' Lemma

Open Questions

Since writing the paper, Farkas' Lemma has gained my attention as the constraint identifying mechanism.

Farkas' Lemma (simplified): Given A and b defining an LP, exactly one of the two proposition it true:

1. $\exists x : Ax = b, x \geq 0,$
2. $\exists y : A^T y \leq 0, b^T y > 0.$

If such a y exists it is a certificate of infeasibility of 1. More importantly, the non-zero entries of y implies variables and constraints which are important for the infeasibility.

Two important features:

If y is a certificate of infeasibility, then so is $\gamma y, \gamma > 0$. This, however, still identifies the same variables and constraints.

An LP can have more than one certificate of infeasibility.

Example.

Continued Application of Farkas' Lemma



$$\dot{x}_1 = -2x_1^3 - 0.5x_1x_2 - 0.5x_1$$

$$\dot{x}_2 = 0.25x_1x_2^2 - 0.125x_1x_2 + 0.25x_2^2 - 0.4125x_2$$

Using $B = [-3, 3]^2$ and $d_V = 2$, the resulting LP is infeasible. Applying Farkas' Lemma once identifies the constraints

$$\mathcal{I}^1 = \{39, 40, 41, 44, 45\}.$$

Setting the corresponding rows of A equal to zero, the reduced problem is also infeasible. Applying Farkas' Lemma on the reduced problem identifies the constraints

$$\mathcal{I}^2 = \{7, 8, 11, 12, 15\}.$$

Setting the corresponding rows of A equal to zero, the second reduced problem is feasible. Thus the union

$$\mathcal{I} = \bigcup_{i=1,2} \mathcal{I}^i = \{7, 8, 11, 12, 15, 39, 40, 41, 44, 45\}$$

contains all the constraints responsible for the infeasibility of the LP.

CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the Bernstein Basis

Lyapunov Functions in the Bernstein Basis

Example

Infeasibility

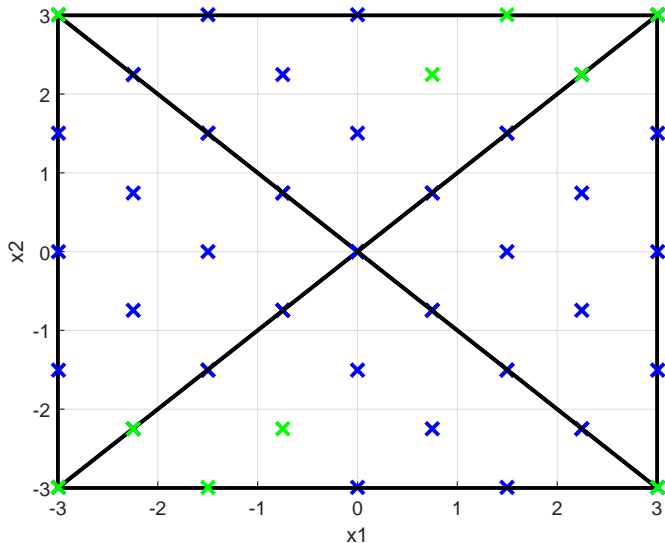
Example Continued

Constraint Identification

26 Farkas' Lemma

Open Questions

Identified Constraints



CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

27 Farkas' Lemma

Open Questions



- ▶ Can the infeasibility certificate be "generalised" to find all constraints at once?
- ▶ How should the information be utilized? What is the "best" way of sub-dividing?
- ▶ Is there any structure in the problem? Actually yes, the A matrix can be set on Dual Block Angular Structure. Dedicated solver?
- ▶ Benchmark Study comparing complexities with other methods.

CortOpt

Tobias Leth

Introduction

Bernstein Basis

Polynomials in the
Bernstein Basis

Lyapunov Functions in
the Bernstein Basis

Example

Infeasibility

Example Continued

Constraint
Identification

Farkas' Lemma

28 Open Questions

Thank you for your attention - Questions?



AALBORG UNIVERSITY
DENMARK