

A Convergent and Efficient Algorithm for Calculating Equilibrium for Chemical Networks of Reversible Binding Reactions

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AMGEN

South San Francisco, California, USA

Society for Industrial and Applied Mathematics

Conference on Applied Algebraic Geometry

**Minisymposium on Algebraic Methods
for Analyzing Biological Interaction Networks**

Georgia Institute of Technology

Atlanta, Georgia, USA

31 July - 4 August 2017

Objective

Solve the equilibrium problem for
complete networks of reversible binding reactions

$$x_j + \sum_{\alpha \in I} \alpha_j a_\alpha x^\alpha = b_j$$

by a *worry-free* algorithm

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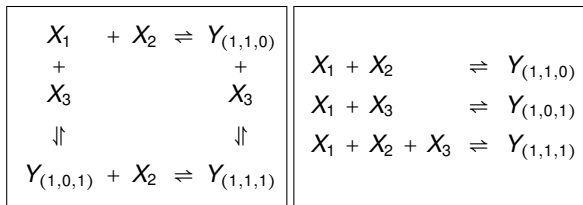
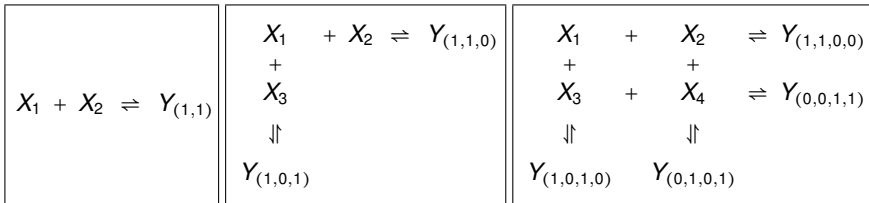
by a *worry-free* algorithm

Requirements for “worry-free”	Weight
Simplicity (yes, this is subjective...)	10%
Computational performance (much less subjective, but still...)	31%
A priori certainty of convergence	59%

Example: Fixed-point iteration of a contraction map

Complete Networks of Reversible Binding Reactions

Examples from Pharmacology in Math Notation

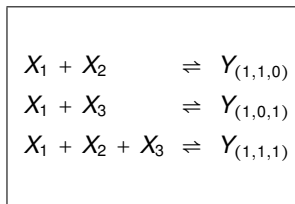
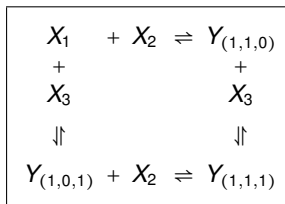


General Idea

- X_1, \dots, X_n : Elementary species (“atoms”)
- Y_α : Composite species of composition α
- Conservation of composition
- Detailed-balance equilibrium

The Equilibrium Problem – Polynomial Formulation

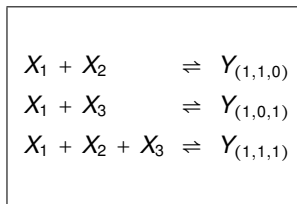
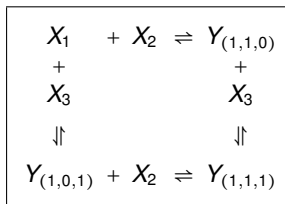
Example: The Allosteric Ternary Complex Model



$$\left\{ \begin{array}{l} X_1 + a_{(1,1,0)} X_1 X_2 + a_{(1,0,1)} X_1 X_3 + a_{(1,1,1)} X_1 X_2 X_3 = b_1 \\ X_2 + a_{(1,1,0)} X_1 X_2 + a_{(1,1,1)} X_1 X_2 X_3 = b_2 \\ X_3 + a_{(1,0,1)} X_1 X_3 + a_{(1,1,1)} X_1 X_2 X_3 = b_3 \end{array} \right.$$

The Equilibrium Problem – Fixed-Point Formulation

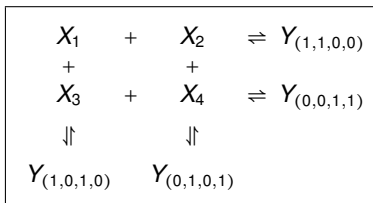
Example: The Allosteric Ternary Complex Model



$$\left\{ \begin{array}{l} x_1 = \frac{b_1}{1 + a_{(1,1,0)} x_2 + a_{(1,0,1)} x_3 + a_{(1,1,1)} x_2 x_3} \\ x_2 = \frac{b_2}{1 + a_{(1,1,0)} x_1 + a_{(1,1,1)} x_1 x_3} \\ x_3 = \frac{b_3}{1 + a_{(1,0,1)} x_1 + a_{(1,1,1)} x_1 x_2} \end{array} \right.$$

The Equilibrium Problem – Polynomial Formulation

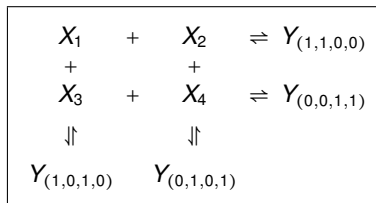
Example: The Receptor-Ligand-Antagonist-Trap Network



$$\begin{cases} X_1 + a_{(1,1,0,0)} X_1 X_2 + a_{(1,0,1,0)} X_1 X_3 & = b_1 \\ X_2 + a_{(1,1,0,0)} X_1 X_2 + a_{(0,1,0,1)} X_2 X_4 & = b_2 \\ X_3 + a_{(1,0,1,0)} X_1 X_3 + a_{(0,0,1,1)} X_3 X_4 & = b_3 \\ X_4 + a_{(0,1,0,1)} X_2 X_4 + a_{(0,0,1,1)} X_3 X_4 & = b_4 \end{cases}$$

The Equilibrium Problem – Fixed-Point Formulation

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The Equilibrium Problem

Polynomial Formulation

$$f(x) = b$$

where

$$f_j(x) = x_j + \sum_{\alpha \in I} \alpha_j a_\alpha x^\alpha$$

Fixed-Point Formulation

$$F(b, x) = x$$

where

$$F_j(b, x) = \frac{b_j}{1 + \sum_{\alpha \in I, \alpha_j \geq 1} \alpha_j a_\alpha x^{\alpha - e_{n,j}}}$$

The Equilibrium Problem – Polynomial Formulation

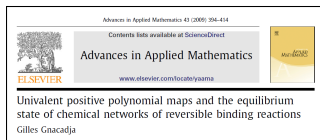
$$f(x) = b$$

where

$$f_i(x) = x_i + \sum_{\alpha \in I} \alpha_i a_\alpha x^\alpha$$

Theorem

The map $f : \mathbb{R}_{\geq 0}^n \rightarrow \mathbb{R}_{\geq 0}^n$ is an infinitely smooth diffeomorphism.



The Equilibrium Problem – Fixed-Point Formulation

$$F(b, x) = x$$

where

$$F_i(b, x) = \frac{b_i}{1 + \sum_{\alpha \in I, \alpha_j \geq 1} \alpha_j a_\alpha x^{\alpha - e_{n,i}}}$$

Theorem

With respect to the metric d , the map $F(b, \cdot) : \mathbb{R}_{>0}^n \rightarrow \mathbb{R}_{>0}^n$ is k -Lipschitz on $]0, b_1] \times \cdots \times]0, b_n]$.

MATHEMATICAL METHODS IN THE APPLIED SCIENCES
Math. Meth. Appl. Sci. 2007; 30:201–211
Published online 23 October 2006 in Wiley InterScience
(www.interscience.wiley.com) DOI: 10.1002/mma.782
MOS subject classification: 47 H 10; 74 G 25; 74 G 30; 74 G 15; 92 C 40



Fixed points of order-reversing maps in $\mathbb{R}_{>0}^n$
and chemical equilibrium

Gilles Gnacadja

The Equilibrium Problem – Fixed-Point Formulation

The metric d and the Lipschitz constant k

$$d(u, v) = \max_{1 \leq i \leq n} |\ln(u_i/v_i)|$$

$$k = \max_{1 \leq i \leq n} k_i$$

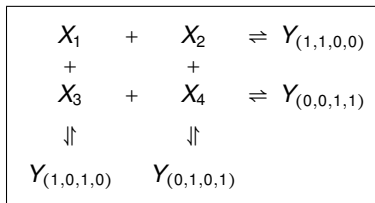
$$k_i = \frac{\sum_{\alpha \in I, \alpha_i \geq 1} |\alpha - e_{n,i}| \alpha_j a_\alpha b^{\alpha - e_{n,i}}}{1 + \sum_{\alpha \in I, \alpha_i \geq 1} \alpha_j a_\alpha b^{\alpha - e_{n,i}}}$$

The Equilibrium Problem – Fixed-Point Formulation

When do we have a contraction?

- ▶ $F(b, \cdot)$ is a contraction if b is small enough.
Not particularly useful for intended application.
- ▶ $F(b, \cdot)$ is a contraction if $(a_\alpha)_{\alpha \in I}$ is small enough.
Even less useful.
- ▶ $F(b, \cdot)$ is a contraction if all composite species are binary.
Useful but narrowly applicable.

Example: The Receptor-Ligand-Antagonist-Trap Network



The Equilibrium Problem – Fixed-Point Formulation

When can we have a contraction?

- ▶ $F(b, \cdot)$ can be “turned into” a contraction if there is only one elementary species.

Not particularly useful for intended application.

Proposition

For a self-map of a convex domain in \mathbb{R}^n , a homotopy with the identity map preserves fixed points.

If $n = 1$ and the original map is monotone-decreasing, then the homotopy transform can be chosen to be a contraction.

The Equilibrium Problem – Fixed-Point Formulation

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Proposition

For a self-map of a convex domain in \mathbb{R}^n , a homotopy with the identity map preserves fixed points.

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Let D be an interval in \mathbb{R} , let ψ a differentiable self-map of D , and suppose there exist $m, M \geq 0$ such that $-M \leq \psi' \leq -m$.

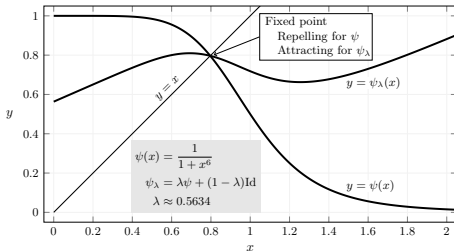
For any $\lambda \in [0, 1]$, let

$\psi_\lambda = \lambda\psi + (1 - \lambda)\text{Id}_D : x \mapsto \lambda\psi(x) + (1 - \lambda)x$ and $k_\lambda = \max(|1 - (1 + m)\lambda|, |1 - (1 + M)\lambda|)$.

The map ψ_λ is k_λ -Lipschitz w.r.t. the ordinary norm.

If (and only if) $0 < \lambda < \frac{2}{1 + M}$, then $k_\lambda < 1$, and iteration of ψ_λ converges to the (unique) fixed point of ψ .

$\arg \min_{\lambda \in [0, 1]} (k_\lambda) = \lambda_0 = \frac{2}{2 + m + M}$; $k_{\lambda_0} = \frac{M - m}{2 + m + M} < 1$.



The Equilibrium Problem – Fixed-Point Formulation

PROBLEM

Can we exploit the properties of $F(b, \cdot)$ to transform it so as to solve the fixed point problem $F(b, x) = x$ by a “worry-free” algorithm?

- ▶ $F(b, \cdot)$ has a unique fixed point in $[0_n, b] \subset \mathbb{R}_{\geq 0}^n$.
- ▶ $F(b, \cdot)$ is order-reversing on $\mathbb{R}_{\geq 0}^n$.
- ▶ We know a Lipschitz constant for $F(b, \cdot)$ on $]0_n, b]$.

The Equilibrium Problem – Fixed-Point Formulation




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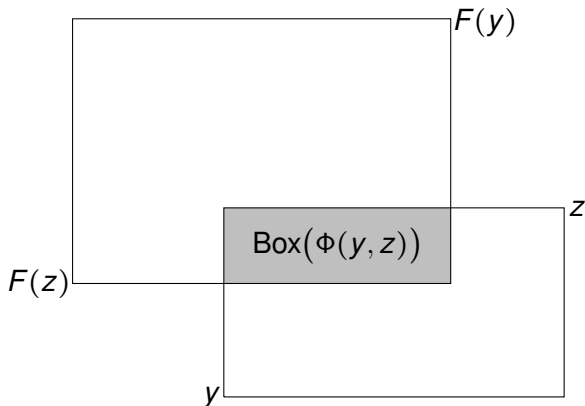
TENTATIVE SOLUTION

Enclosure algorithm

Requirements for “worry-free”	Weight
 Simplicity	10%
 Computational performance	31%
 A priori certainty of convergence	59%

Enclosure Algorithm – Fixed-Box Iteration

$$\Phi(y, z) := (\sup(y, F(z)), \inf(z, F(y)))$$



$$\begin{aligned} \text{Fix}(F) \cap \text{Box}(y, z) &= \text{Fix}(F) \cap \text{Box}(\Phi(y, z)) \\ &= \text{Fix}(F) \cap \text{Box}(\Phi^k(y, z)), \quad \forall k \in \mathbb{Z}_{\geq 0} \end{aligned}$$

Enclosure Algorithm – Stopping Fixed-Box Iteration

Descending sequence $(\text{Box}(\Phi^k(y, z)))_{k \in \mathbb{Z}_{\geq 0}}$ converges to $\text{Box}(\Phi^\infty(y, z)) := \bigcap_{k \in \mathbb{Z}_{\geq 0}} \text{Box}(\Phi^k(y, z))$.

- ▶ $\text{Box}(\Phi^k(y, z)) = \emptyset$ for some $k \in \mathbb{Z}_{\geq 0}$:
Discard $\text{Box}(y, z)$
- ▶ For some $k \in \mathbb{Z}_{\geq 0}$, $\text{Box}(\Phi^k(y, z))$ is found to contain a good approximation of the fixed point :
Done
- ▶ Otherwise :
Subdivide $\text{Box}(\Phi^{k_{\max}}(y, z))$ and repeat

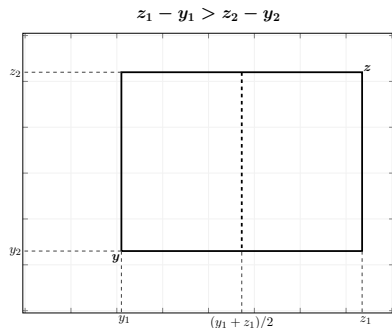
Enclosure Algorithm – Subdivision Strategy

There are numerous box subdivision strategies.

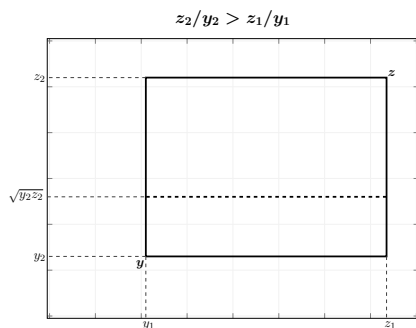
Best (relatively) performance achieved with
geometric long-edge bisection:

Cut box orthogonally to the first multiplicatively-longest edge at the geometric center.

Arithmetic



Geometric



Enclosure Algorithm – Starting Box

▶ $F^2(0) \leq F^4(0) \leq F^6(0) \leq \dots \leq F^5(0) \leq F^3(0) \leq F(0)$

▶ $(F^k(0))_{k \in \mathbb{Z}_{\geq 0}}$ converges:

Done

▶ $(F^k(0))_{k \in \mathbb{Z}_{\geq 0}}$ accumulates to a 2-orbit :

Perform enclosure algorithm starting at

Box($F^{2h_{\max}}(0), F^{2h_{\max}-1}(0)$)

Evolution of the algorithm on a 4-dimensional example

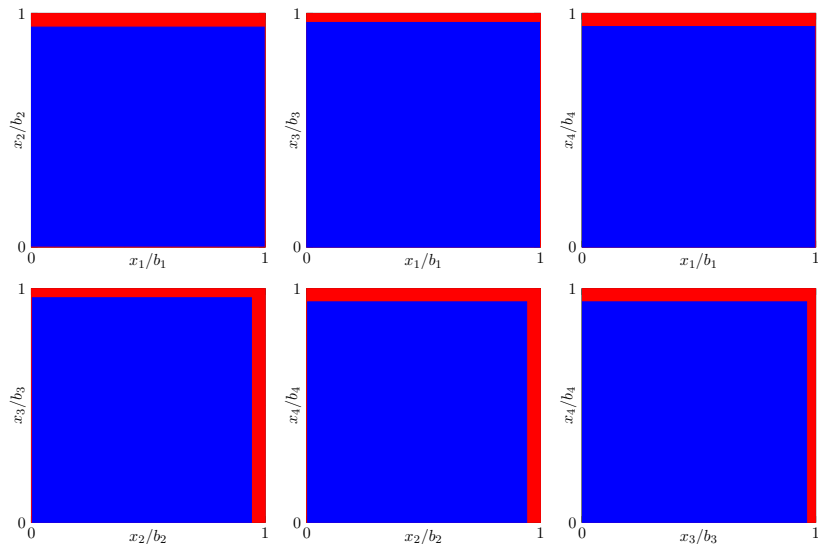
$$\begin{aligned}X_1 + X_2 &\rightleftharpoons Y_{(1,1,0,0)} \\X_1 + X_3 &\rightleftharpoons Y_{(1,0,1,0)} \\X_2 + X_3 &\rightleftharpoons Y_{(0,1,1,0)} \\X_3 + X_4 &\rightleftharpoons Y_{(0,0,1,1)} \\Y_{(1,0,1,0)} + X_4 &\rightleftharpoons Y_{(1,0,1,1)}\end{aligned}$$

$$\left\{ \begin{aligned}f_1(x) &= x_1 + a_{(1,1,0,0)}x_1x_2 + a_{(1,0,1,0)}x_1x_3 + a_{(1,0,1,1)}x_1x_3x_4 \\f_2(x) &= x_2 + a_{(1,1,0,0)}x_1x_2 + a_{(0,1,1,0)}x_2x_3 \\f_3(x) &= x_3 + a_{(0,1,1,0)}x_2x_3 + a_{(1,0,1,0)}x_1x_3 + a_{(0,0,1,1)}x_3x_4 + a_{(1,0,1,1)}x_1x_3x_4 \\f_4(x) &= x_4 + a_{(0,0,1,1)}x_3x_4 + a_{(1,0,1,1)}x_1x_3x_4\end{aligned}\right.$$

$$\left\{ \begin{aligned}F_1(b, x) &= \frac{b_1}{1 + a_{(1,1,0,0)}x_2 + a_{(1,0,1,0)}x_3 + a_{(1,0,1,1)}x_3x_4} \\F_2(b, x) &= \frac{b_2}{1 + a_{(1,1,0,0)}x_1 + a_{(0,1,1,0)}x_3} \\F_3(b, x) &= \frac{b_3}{1 + a_{(0,1,1,0)}x_2 + a_{(1,0,1,0)}x_1 + a_{(0,0,1,1)}x_4 + a_{(1,0,1,1)}x_1x_4} \\F_4(b, x) &= \frac{b_4}{1 + a_{(0,0,1,1)}x_3 + a_{(1,0,1,1)}x_1x_3}\end{aligned}\right.$$

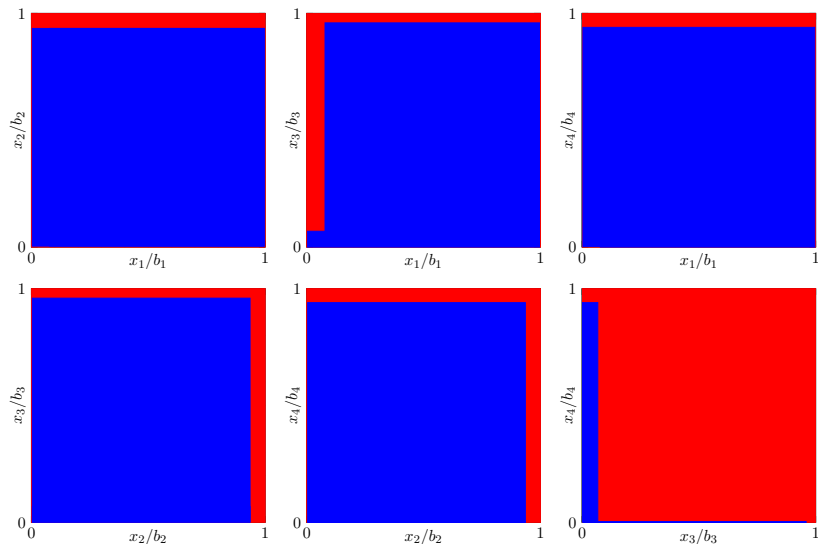
Evolution of the algorithm on a 4-dimensional example

Starting box from stalled fixed-point iteration



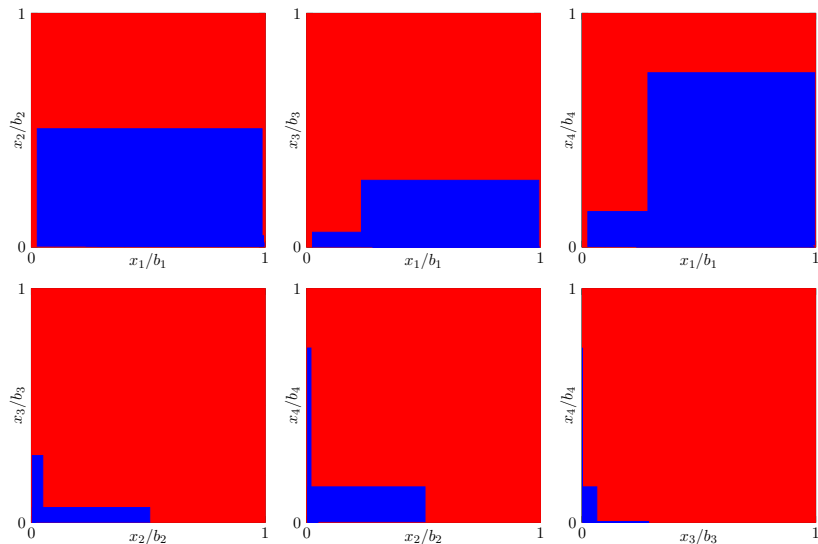
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 1 of 18 – Boxes examined/admitted/discarded : 2/2/0



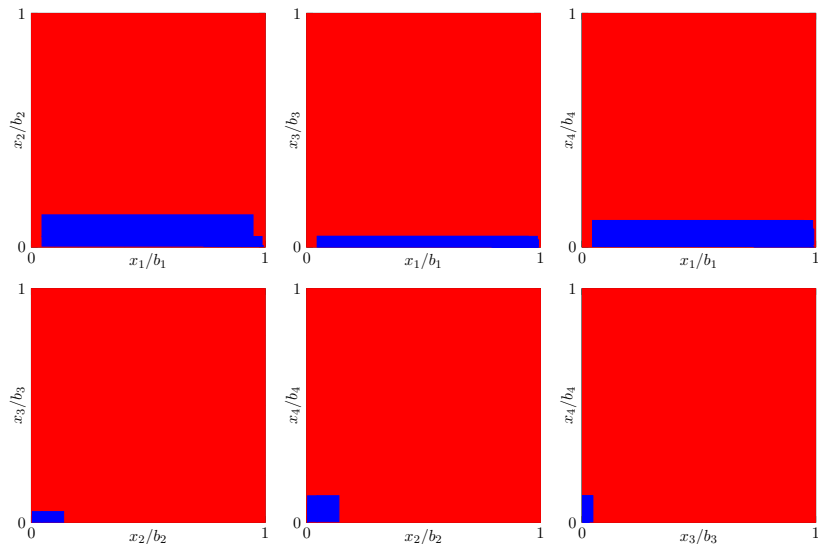
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 2 of 18 – Boxes examined/admitted/discarded : 4/3/1



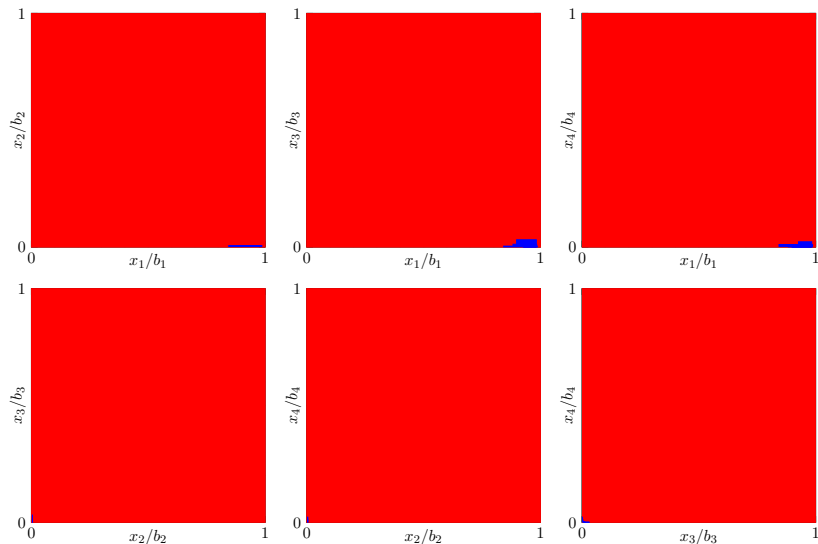
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 3 of 18 – Boxes examined/admitted/discarded : 6/4/2



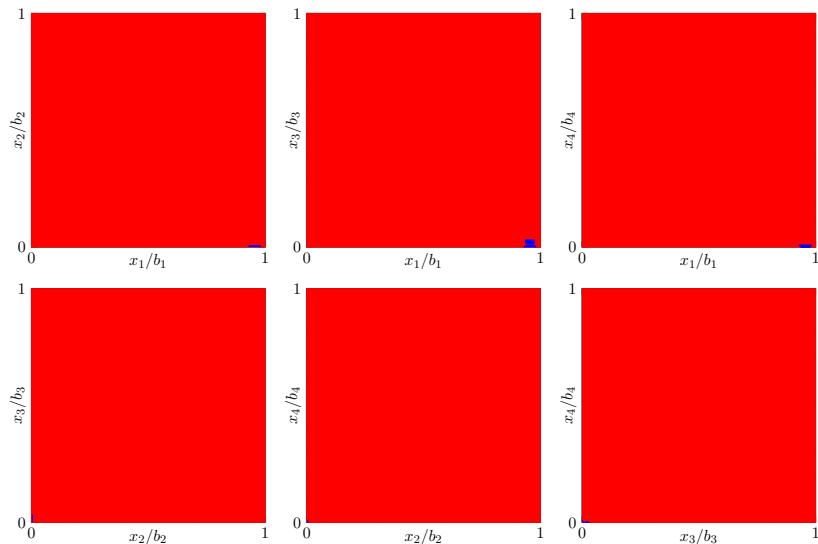
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 4 of 18 – Boxes examined/admitted/discarded : 8/4/4



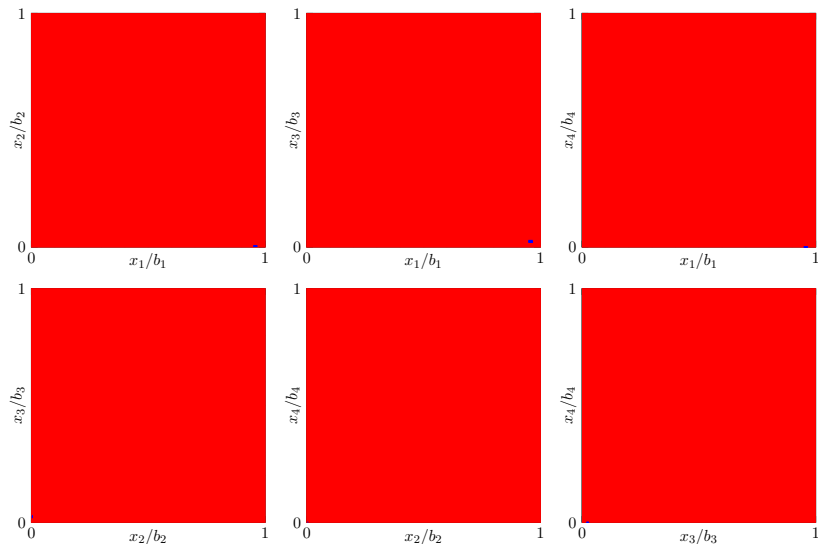
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 5 of 18 – Boxes examined/admitted/discarded : 8/4/4



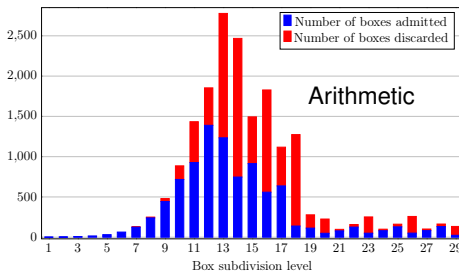
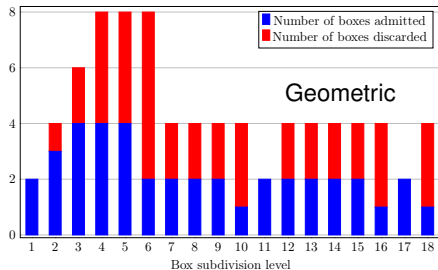
Evolution of the algorithm on a 4-dimensional example

Geometric long-edge bisection – Level 6 of 18 – Boxes examined/admitted/discarded : 8/2/6 [END]



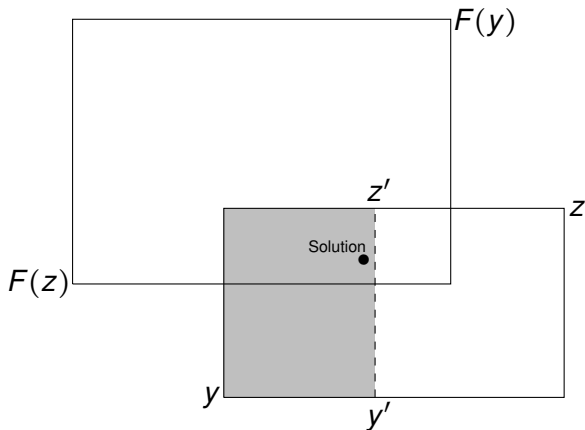
Evolution of the algorithm on a 4-dimensional example

Boxes admitted/discarded : geometric-vs-arithmetic long-edge bisection



Another Approach: Box Squeezing

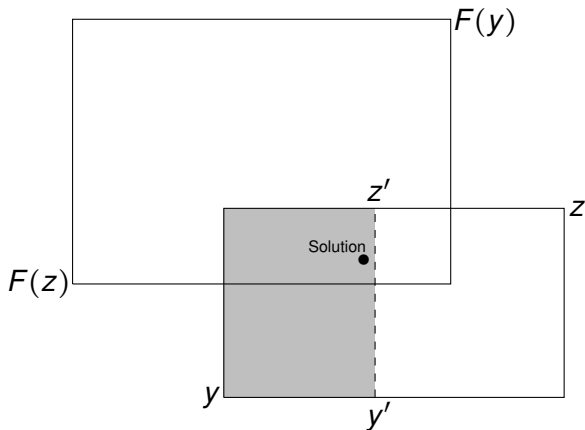
Fixed-point formulation optional. Surprisingly unimpressive performance.



$$\begin{cases} y \leq F(y) \\ F(z) \leq z \end{cases} \rightarrow \rightarrow \rightarrow \begin{cases} y' \not\leq F(y') \\ F(z') \leq z' \end{cases}$$

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Fixed-point formulation optional. Surprisingly unimpressive performance.



$$\begin{cases} y \leq F(y) \\ F(z) \leq z \end{cases} \quad \dashrightarrow \dashrightarrow \quad \begin{cases} y' \not\leq F(y') \\ F(z') \leq z' \end{cases}$$

$$\begin{cases} f(y) \leq b \\ b \leq f(z) \end{cases} \quad \dashrightarrow \dashrightarrow \quad \begin{cases} f(y') \not\leq b \\ b \leq f(z') \end{cases}$$

Recapitulation

$$f_j(x) = x_j + \sum_{\alpha \in I} \alpha_j a_\alpha x^\alpha$$

$$F_j(b, x) = \frac{b_j}{1 + \sum_{\alpha \in I, \alpha_j \geq 1} \alpha_j a_\alpha x^{\alpha - e_{n,j}}}$$

KEY PROPERTIES

- ▶ $f : \mathbb{R}_{\geq 0}^n \rightarrow \mathbb{R}_{\geq 0}^n$ is an infinitely smooth diffeomorphism.
- ▶ $F(b, \cdot)$ has a unique fixed point in $]0_n, b] \subset \mathbb{R}_{\geq 0}^n$.
- ▶ $F(b, \cdot)$ is order-reversing on $\mathbb{R}_{\geq 0}^n$.
- ▶ We know a Lipschitz constant for $F(b, \cdot)$ on $]0_n, b]$.

PROBLEM

Solve $\mathbf{F}(\mathbf{b}, \mathbf{x}) = \mathbf{x}$ (or $\mathbf{f}(\mathbf{x}) = \mathbf{b}$) by a worry-free algorithm.

WORRY-FREE ALGORITHM

Simplicity	10%
Computational performance	31%
A priori certainty of convergence	59%