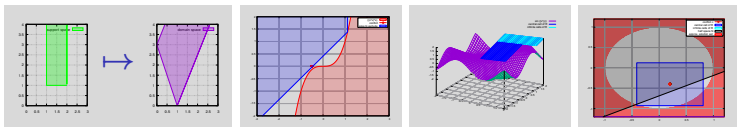


Generalizing Taylor models for multivariate real functions [×]



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- 1 Linearizing non-linearity
- 2 Notes on Interval arithmetic
- 3 Representing multivariate functions
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Combine **reliable** real computations and **resolution** in a CDCL-style calculus called **KSMT** aiming at SMT solvers for **nonlinear problems**.

Main parts:

- 1 use CNF \mathcal{C} in **separated linear form** $\mathcal{C} = \mathcal{L} \wedge \mathcal{N}$
 - ▶ \mathcal{L} – clauses of linear inequalities: $q_1x_1 + q_2x_2 + \dots + q_nx_n + q_0 \diamond 0$

$$\begin{array}{r} 2x_1 - 4x_2 - 2x_3 - 2 > 0 \\ \vee \quad \quad \quad 4x_2 + 2x_3 + 1 \geq 0 \end{array}$$

- ▶ \mathcal{N} – **unit clauses** of non-linear inequalities : $f(\bar{x}) \diamond y$

$$\begin{array}{r} x_2^2 \cdot x_1 \leq x_3 \\ \sin(x_1^2 + x_2^2) > 0 \end{array}$$

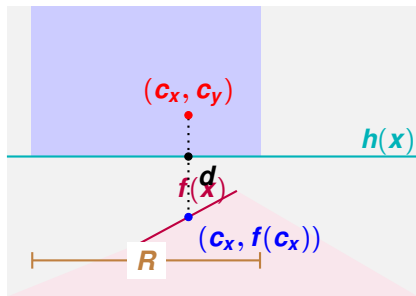
- 2 conflict-driven calculus with well-known steps of
 - ▶ (R): linear inequality resolution
 - ▶ (B): backjumps
 - ▶ (A): assignment refinement
- 3 new step (L): **local linearisations** for resolving **non-linear conflicts**

Constraint $\mathbf{P} \equiv \mathbf{y} \leq \mathbf{f}(\mathbf{x})$, assignment $\mathbf{x} \mapsto \mathbf{c}_x$, $\mathbf{y} \mapsto \mathbf{c}_y$

Interval linearisation:

1. check for **conflict**
2. choose intermediate \mathbf{d}
3. use **constant function h** with $h(\mathbf{c}_x) = \mathbf{d}$
4. and find suitable **polytope(s) R** with

$$\mathbf{y} \leq \mathbf{f}(\mathbf{x}) \Rightarrow (\mathbf{x} \notin R \vee \mathbf{y} \leq \mathbf{h}(\mathbf{x}))$$



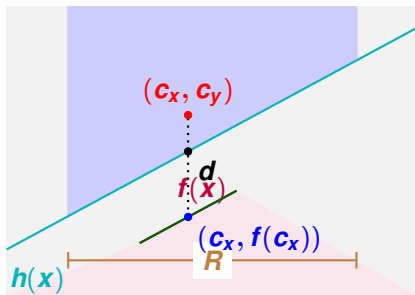
reasonable: R large with $R \subseteq \{\mathbf{x} \mid \mathbf{h}(\mathbf{x}) \geq \mathbf{f}(\mathbf{x})\}$

Constraint $P \equiv y \leq f(x)$, assignment $x \mapsto c_x, y \mapsto c_y$

Tangent space linearisation:

1. check for **conflict**
2. choose intermediate d
3. approximate **slope** ∂
4. choose **linear function** h with $h(c_x) = d$
5. and find suitable **polytope(s)** R with

$$y \leq f(x) \Rightarrow (x \notin R \vee y \leq h(x))$$



reasonable: R large with $R \subseteq \{x \mid h(x) \geq f(x)\}$

Example:

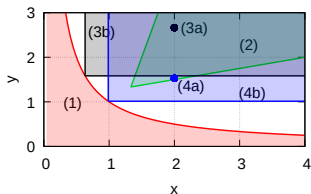
$$C = (y \leq 1/x)$$

$$\wedge (y \geq x/4 + 1)$$

$$\wedge (y \leq 4 \cdot (x - 1))$$

$$\wedge ((x \leq \frac{12}{19}) \vee (y \leq \frac{19}{12}))$$

$$\wedge ((x \leq \frac{220}{223}) \vee (y \leq \frac{223}{220}))$$



Linearisation of $(y \leq 1/x)$
and conflict (c_x, c_y) :

- use $d := (c_y + 1/c_x)/2$

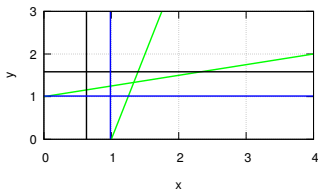
- $(y \leq 1/x) \rightarrow$

$$((x \leq 1/d) \vee (y \leq d))$$

rule	α	note
(A)	$x \mapsto 2$	
(A)	$x \mapsto 2, y \mapsto \frac{8}{3}$	(3a)
(L)	$x \mapsto 2, y \mapsto \frac{3}{2}$	(3b)
(B)	$x \mapsto 2$	
(A)	$x \mapsto 2, y \mapsto \frac{84}{55}$	(4a)
(L)	$x \mapsto 2, y \mapsto \frac{54}{55}$	(4b)
	...	
n/a		unsat

Interpretation:

$$\begin{aligned} \mathcal{C} = & (y \leq 1/x) \\ & \wedge (y \geq x/4 + 1) \\ & \wedge (y \leq 4 \cdot (x - 1)) \\ & \wedge ((x \leq \frac{12}{19}) \vee (y \leq \frac{19}{12})) \\ & \wedge ((x \leq \frac{220}{223}) \vee (y \leq \frac{223}{220})) \end{aligned}$$



- lin. predicates \sim prop. variables

$$\begin{aligned} \mathcal{C} = & N \\ & \wedge L_1 \\ & \wedge L_2 \\ & \wedge (L_3 \vee L_4) \\ & \wedge (L_5 \vee L_6) \end{aligned}$$

- Prop. variables:
 $\{N, L_1, L_2\} \cup \{L_3, L_4, L_5, L_6\}$
- linearisations \sim clauses
 $(L_3 \vee L_4), (L_5 \vee L_6)$
- parallelity \sim implications
 $(\neg L_3 \vee L_5), (\neg L_6 \vee L_4)$

- adding propositional variables is expensive
- adding clauses is quite cheap

Requirements (simplified...) for computing linearisations:

- input:

- ▶ function f , defined by term t using basic functions
- ▶ vector $\mathbf{c}_x \in \mathbb{Q}^d$ and value $\mathbf{c}_y \in \mathbb{Q}$ with $\mathbf{c}_y > f(\mathbf{c}_x)$
- ▶ set $\mathbf{G} = \{\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_n\}$ of linear functions

defining initial polytope $\mathbf{R} := \bigcap_{i \leq n} \{\bar{\mathbf{x}} \in \mathbb{R}^d \mid \mathbf{g}_i(\bar{\mathbf{x}}) > \mathbf{0}\}$ with $\mathbf{c}_x \in \mathbf{R}$

- output:

- ▶ function h and set $\mathbf{G}' = \{\mathbf{g}'_1, \mathbf{g}'_2, \dots, \mathbf{g}'_m\}$ (all linear)

defining a polytope $\mathbf{R}' := \bigcap_{i \leq m} \{\bar{\mathbf{x}} \in \mathbb{R}^d \mid \mathbf{g}'_i(\bar{\mathbf{x}}) > \mathbf{0}\}$ with $\mathbf{c}_x \in \mathbf{R}'$

and

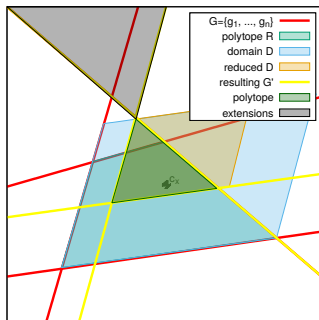
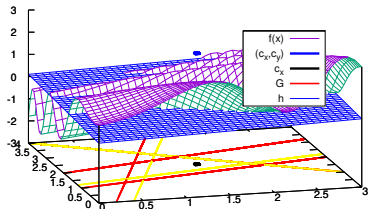
$$\mathbf{c}_y > h(\mathbf{c}_x), \quad \forall \bar{\mathbf{x}} \in \mathbf{R}' \quad h(\bar{\mathbf{x}}) > f(\bar{\mathbf{x}})$$

and (possibly) additional clauses for cells \mathbf{R}'' satisfying
 $\forall \bar{\mathbf{x}} \in \mathbf{R}'' \quad h(\bar{\mathbf{x}}) > f(\bar{\mathbf{x}})$

- input and output in symbolical form, only using rational numbers
- preferably with 'large' overlap $\mathbf{G}' \cap \mathbf{G}$

Basic algorithm

- Given: f , \mathbf{c}_x , \mathbf{c}_y , \mathbf{G}
- $\mathbf{R} := \bigcap_{i \leq n} \{\bar{\mathbf{x}} : \mathbf{g}_i(\bar{\mathbf{x}}) > 0\}$
- construct 'simple' \mathbf{D} with $\mathbf{c}_x \in \mathbf{D}$
- repeat
 - compute linear function h
 - with $(\forall \bar{\mathbf{x}} \in \mathbf{D}) h(\bar{\mathbf{x}}) > f(\bar{\mathbf{x}})$
 - if $\mathbf{c}_y \leq h(\mathbf{c}_x)$,
 - shrink \mathbf{D} , keeping $\mathbf{c}_x \in \mathbf{D}$
- until $\mathbf{c}_y > h(\mathbf{c}_x)$
- match \mathbf{D} with $\{\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_n\}$
- return h and $\mathbf{G}' = \{\mathbf{g}'_1, \dots\}$
- return polytopes with $h(\bar{\mathbf{x}}) > f(\bar{\mathbf{x}})$



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Definition 2.1

- $[a:b]$ is the closed interval $\{z \in \mathbb{R} \mid a \leq z \leq b\}$
- $[a]$ is the point interval $\{a\} = [a:a]$
- Infinite intervals: $[a:\infty]$, $[-\infty:b]$, $[-\infty:\infty]$
- Let \mathbb{K} be a subset of \mathbb{R} , e.g. $\mathbb{Q}, \mathbb{R}_c, \mathbb{D}$.

The *extended real intervals* \mathbb{IK} with *bounds from* \mathbb{K} are

$$\begin{aligned} \mathbb{IK} = & \{ [a:b] \mid a, b \in \mathbb{K}, a \leq b \} \\ & \cup \{ [a : \infty] \mid a \in \mathbb{K} \} \\ & \cup \{ [-\infty : b] \mid b \in \mathbb{K} \} \\ & \cup \{ [-\infty : \infty] \} \end{aligned}$$

Usual operations (perhaps approximate) on intervals:

- standard arithmetic ($+$, $-$, \cdot , $/$), flags for finiteness
- \inf , \sup , mid , rad , sgn
- constructing point intervals, rational approximations

Valid laws for $\mathbf{A}, \mathbf{B}, \mathbf{C} \in \mathbb{IK}$, $\mathbf{a} \in \mathbb{K}$, if \mathbb{K} is a ring:

- $\mathbf{A} + \mathbf{B} = \mathbf{B} + \mathbf{A}$ and $\mathbf{A} \cdot \mathbf{B} = \mathbf{B} \cdot \mathbf{A}$, commutative
- $\mathbf{A} \subseteq \mathbf{C}; \mathbf{B} \subseteq \mathbf{D} \Rightarrow \mathbf{A} \circ \mathbf{B} \subseteq \mathbf{C} \circ \mathbf{D}$ inclusion monotone
- $(\mathbf{A} + \mathbf{B}) \cdot \mathbf{C} \subseteq \mathbf{A} \cdot \mathbf{C} + \mathbf{B} \cdot \mathbf{C}$, subdistributive
- $(\mathbf{A} + [\mathbf{b}]) \cdot [\mathbf{c}] = \mathbf{A} \cdot [\mathbf{c}] + [\mathbf{b}] \cdot \mathbf{c}$, distributive for point intervals

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main task: Given f and \bar{x} , find

- linear function h and polytope D with $\bar{x} \in D$
- with $\forall \bar{x} \in D : h(\bar{x}) > f(\bar{x})$

intermediate goal:

- data type for approximations of functions on polytopes
- interface to data type:
 - ▶ pointwise application of important real operators like ADD, SUB, MULT, SIN, ...

$$\begin{aligned}\text{ADD}(f, g) = h &\rightsquigarrow \forall \bar{x} \in D : h(\bar{x}) = f(\bar{x}) + g(\bar{x}) \\ \text{SIN}(f) = h &\rightsquigarrow \forall \bar{x} \in D : h(\bar{x}) = \sin(f(\bar{x}))\end{aligned}$$

- ▶ (approximative) evaluation at real arguments
- ▶ linear bounds on (possibly infinite) polytopes

(\rightsquigarrow Representations in TTE sense can be derived...)

Generalize the idea of **Taylor models** [Makino/Berz]:

- use multivariate polynomial approximations
- with intervals coefficients
- use vectors $\bar{\lambda} = (\lambda_1, \dots, \lambda_k)^T$ of parameters $\lambda_i \in \mathbf{S}_i$
(‘error symbols’ in the setting of Taylor models)

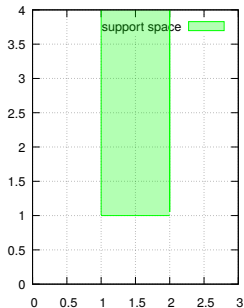
Modifications:

- use polynomials with (possibly infinite) interval coefficients
(instead of a single finite interval for all errors)
- use rational / dyadic / computable real numbers
(instead of double precision)
- use (possibly infinite) support space $\mathbf{S} = \mathbf{S}_1 \times \dots \times \mathbf{S}_k$
(instead of ‘usual’ hypercube $[-1, 1]^k$)
- use parallelotopes as domain space \mathbf{D}
via a linear transformation Δ applied to \mathbf{S}
(instead of axes-aligned boxes)

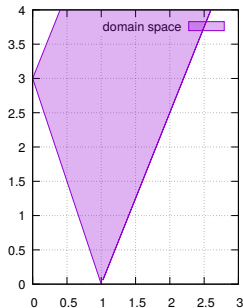
Example: Consider

$$\mathbf{S} = \begin{pmatrix} [1 : 2] \\ [1 : \infty) \end{pmatrix}, \mathbf{A} = \begin{pmatrix} 1 & 2 \\ -3 & 5 \end{pmatrix}, \mathbf{B} = \begin{pmatrix} -3 \\ 1 \end{pmatrix}, \mathbf{A}^{-1} = \begin{pmatrix} \frac{5}{11} & \frac{-2}{11} \\ \frac{3}{11} & \frac{1}{11} \end{pmatrix}$$

support \mathbf{S} and domain \mathbf{D} are as follows:



$$\begin{aligned} \Delta(\bar{\mathbf{x}}) &= \mathbf{A} \cdot \bar{\mathbf{x}} + \mathbf{B} \\ &\longrightarrow \\ \mathbf{S} \ni \bar{\mathbf{x}} & \qquad \bar{\mathbf{x}} \in \mathbf{D} \\ &\longleftarrow \\ \Delta^{-1}(\mathbf{x}) &= \mathbf{A}^{-1} \cdot (\mathbf{x} - \mathbf{B}) \end{aligned}$$



Definition 3.2

- A *Taylor model base (TMB)* \mathbf{T} consists of the following components:
 - ▶ a column vector $\bar{\mathbf{S}} = (\mathbf{S}_1, \dots, \mathbf{S}_k)^T$ with $\mathbf{S}_i \in \mathbb{K}$
 - ▶ an invertible affine transformation Δ ,
given by a nonsingular matrix $\mathbf{A} \in \mathbb{K}^{k \times k}$ and a vector $\mathbf{B} \in \mathbb{K}^k$
- The *support space* $\mathbf{S}_T \subseteq \mathbb{R}^k$ of \mathbf{T} is the cartesian product

$$\mathbf{S}_T = \times \mathbf{S}_i$$

- The *domain space* $\mathbf{D}_T \subseteq \mathbb{R}^k$ of \mathbf{T} is the set

$$\mathbf{D}_T = \{\mathbf{A} \cdot \bar{\lambda} + \mathbf{B} \mid \bar{\lambda} \in \mathbf{S}_T\}$$

Definition 3.3

- Consider multivariate polynomials \mathbf{p} in $\bar{\lambda}$:

$$\mathbf{p}(\bar{\lambda}) = \sum_{\bar{n}} \mathbf{c}_{\bar{n}} \cdot \bar{\lambda}^{\bar{n}}$$

where \mathbf{p} is given by its coefficients $\mathbf{c}_{\bar{n}} \in \mathbb{IK}$.

- A function $\mathbf{f} : \subseteq \mathbb{R}^k \rightarrow \mathbb{R}$ is approximated by \mathbf{p} w.r.t. \mathbf{T} , iff

$$\mathbf{D}_{\mathbf{T}} \subseteq \text{dom}(\mathbf{f})$$

and

$$\forall \bar{\mathbf{x}} \in \mathbf{D}_{\mathbf{T}} \quad \mathbf{f}(\bar{\mathbf{x}}) \in \mathbf{p}(\Delta^{-1}(\bar{\mathbf{x}}))$$

(in symbols: $\mathbf{f} \sim^{\mathbf{T}} \mathbf{p}$)

Convergence of the approximations:

Lemma 3.4 (Weierstraß approximation theorem)

If \mathbf{D}_T is compact, every continuous $\mathbf{f} : \mathbf{D}_T \rightarrow \mathbb{R}$ can be approximated arbitrarily precise by polynomials.

Lemma 3.5 (Interval arithmetic)

For every continuous $\mathbf{f} : \mathbb{R}^k \rightarrow \mathbb{R}$, every $\bar{\mathbf{x}} \in \mathbb{R}^k$ and for any precision, there are T and a linear \mathbf{p} with $\bar{\mathbf{x}} \in \mathbf{D}_T$ approximating \mathbf{f} on \mathbf{D}_T with the desired precision.

So, choose between extremes of high order polynomials or small domains...

Interval oriented operations: Replace occurrence of λ_i in \mathbf{p} by \mathbf{s}_i , e.g.

- **Sweep**: Replace $\bar{\lambda}^{\bar{n}}$ within \mathbf{p} by $\bar{\mathbf{S}}^{\bar{n}}$:

$$\dots + \underbrace{\mathbf{c}}_{\text{interval}} \cdot \underbrace{\bar{\lambda}^{\bar{n}} \cdot \bar{\lambda}^{\bar{m}}}_{\text{monomial}} + \dots \quad \rightsquigarrow \quad \dots + \underbrace{\mathbf{c} \cdot \bar{\mathbf{S}}^{\bar{n}}}_{\text{interval}} \cdot \underbrace{\bar{\lambda}^{\bar{m}}}_{\text{monomial}} + \dots$$

- **Reduce**: Sweep *all* monomials $\bar{\lambda}^{\bar{n}}$ in \mathbf{p} with $\bar{n} \neq \bar{0}$ in \mathbf{p} .

$$\mathbf{p} \quad \rightsquigarrow \quad \mathbf{p}\left(\begin{pmatrix} \mathbf{S}_1 \\ \vdots \\ \mathbf{S}_k \end{pmatrix}\right) \in \mathbb{IK}$$

- **Pointifying**: For arbitrary *point* \mathbf{d} apply

$$\dots + \underbrace{\mathbf{c}}_{\text{interval}} \cdot \underbrace{\bar{\lambda}^{\bar{n}}}_{\text{monomial}} + \dots \quad \rightsquigarrow \quad \dots + \underbrace{\mathbf{d}}_{\text{point}} \cdot \underbrace{\bar{\lambda}^{\bar{n}}}_{\text{monomial}} + \underbrace{(\mathbf{c} - \mathbf{d}) \cdot \bar{\mathbf{S}}^{\bar{n}}}_{\text{interval}} + \dots$$

Further operations on the Taylor models, e.g.

- **Splitting**: Replace interval coefficient \mathbf{c} by a *new* variable $\lambda_{\mathbf{c}}$ with support \mathbf{c} , e.g.

$$\dots + \underbrace{\mathbf{c}}_{\text{interval}} \cdot \underbrace{\bar{\lambda}^{\bar{n}}}_{\text{monomial}} + \dots \quad \rightsquigarrow \quad \dots + \underbrace{[1]}_{\text{interval}} \cdot \underbrace{\lambda_{\mathbf{c}} \cdot \bar{\lambda}^{\bar{n}}}_{\text{monomial}} + \dots$$

$\lambda_{\mathbf{c}}$: shortcut for this occurrence of interval \mathbf{c} in further computations, without changes to domain and support spaces!

- **Square sweep**: Preferably replace λ_i^2 by $\mathbf{S}_i^2 \subseteq \mathbb{R}_0^+$ (instead of $\mathbf{S}_i \cdot \mathbf{S}_i \subseteq \mathbb{R}$) in case $\mathbf{0} \in \mathbf{S}_i$.

Lemma 3.6

Suppose $\mathbf{p} \rightsquigarrow \mathbf{p}'$ via sweeping/reducing/pointifying/splitting, then

$$\mathbf{f} \sim^T \mathbf{p} \quad \Longrightarrow \quad \mathbf{f} \sim^T \mathbf{p}'$$

- Reducing \mathbf{p} leads to a single interval $\mathbf{c} \in \mathbb{IK}$.
So if $\mathbf{0} \notin \mathbf{c}$, then $\mathbf{f}(\bar{\mathbf{x}}) \neq \mathbf{0}$ on \mathbf{D}_T .
- Sweep and pointify \mathbf{p} leading to

$$\mathbf{p}' = \mathbf{c} + \bar{\mathbf{d}}_{\bar{n}} \cdot \bar{\lambda}^{\bar{n}}$$

with *interval* \mathbf{c} and *point vectors* $\bar{\mathbf{d}}_{\bar{n}}$.

Then, if \mathbf{c} has finite upper bound $\sup(\mathbf{c})$,

$$\mathbf{h}(\bar{\mathbf{x}}) := \sup(\mathbf{c}) + \bar{\mathbf{d}}_{\bar{n}} \cdot (\Delta^{-1}(\bar{\mathbf{x}}))^{\bar{n}}$$

is a *polynomial* with

$$\forall \bar{\mathbf{x}} \in \mathbf{D}_T \quad \mathbf{h}(\bar{\mathbf{x}}) \geq \mathbf{f}(\bar{\mathbf{x}})$$

(similar for lower bound $\inf(\mathbf{c})$)

Example 3.7

The projections $\mathbf{pr}_i: \mathbf{D}_T \rightarrow \mathbb{R}, \begin{pmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_k \end{pmatrix} \mapsto \mathbf{x}_i$ are approximated by

$$\overline{\mathbf{pr}}_i: \mathbf{S}_T \rightarrow \mathbb{R}, \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_k \end{pmatrix} \mapsto \mathbf{e}_i \cdot (\mathbf{A} \cdot \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_k \end{pmatrix} + \mathbf{B})$$

where $\mathbf{e}_i = (\underbrace{[\mathbf{0}], \dots, [\mathbf{0}]}_i, [\mathbf{1}], [\mathbf{0}], \dots, [\mathbf{0}])$ is the i -th unit vector consisting of point intervals.

In example with $\mathbf{A} = \begin{pmatrix} 1 & 2 \\ -3 & 5 \end{pmatrix}, \mathbf{B} = \begin{pmatrix} -3 \\ 1 \end{pmatrix}$

$$\begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} \mapsto \mathbf{x}_1 \quad \sim^T \quad \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} \mapsto [\mathbf{1}] \cdot \lambda_1 + [\mathbf{2}] \cdot \lambda_2 + [-\mathbf{3}]$$
$$\begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} \mapsto \mathbf{x}_2 \quad \sim^T \quad \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} \mapsto [-\mathbf{3}] \cdot \lambda_1 + [\mathbf{5}] \cdot \lambda_2 + [\mathbf{1}]$$

Example 3.8

Let $\mathbf{f}_1 \sim^T \mathbf{p}_1$ and $\mathbf{f}_2 \sim^T \mathbf{p}_2$.

Let $\begin{Bmatrix} \circ \\ \diamond \end{Bmatrix}$ be $\begin{Bmatrix} \text{pointwise} \\ \text{formal} \end{Bmatrix}$ product of $\begin{Bmatrix} \text{functions} \\ \text{polynomials} \end{Bmatrix}$.

Then $\mathbf{f}_1 \circ \mathbf{f}_2 \sim^T \mathbf{p}_1 \diamond \mathbf{p}_2$. The same holds for sum and difference.

In example with $\mathbf{A} = \begin{pmatrix} 1 & 2 \\ -3 & 5 \end{pmatrix}$, $\mathbf{B} = \begin{pmatrix} -3 \\ 1 \end{pmatrix}$

$\begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} \mapsto \mathbf{x}_1 \cdot \mathbf{x}_2$ is approximated by

$$\begin{aligned} \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} &\mapsto ([1]\lambda_1 + [2]\lambda_2 + [-3]) \cdot ([-3]\lambda_1 + [5]\lambda_2 + [1]) \\ &= [-3]\lambda_1^2 + [-1]\lambda_1\lambda_2 + [10]\lambda_2^2 + [10]\lambda_1 + [-13]\lambda_2 + [-3] \end{aligned}$$

Consider example $f(\mathbf{x}_1, \mathbf{x}_2) = \mathbf{x}_1 \cdot \mathbf{x}_2$ with TMB as above, using $\mathbf{S}_1 = [1:2]$ and $\mathbf{S}_2 = [1:\infty]$:

$$\begin{aligned}
 & [-3]\lambda_1^2 + [-1]\lambda_1\lambda_2 + [10]\lambda_2^2 + [10]\lambda_1 + [-13]\lambda_2 + [-3] \\
 \xrightarrow{\text{sweep}} & [-3] \cdot [1:2] \cdot \lambda_1 + [-1][1:2] \cdot \lambda_2 + [10] \cdot [1:\infty] \cdot \lambda_2 \\
 & \quad + [10]\lambda_1 + [-13]\lambda_2 + [-3] \\
 = & [4:7] \cdot \lambda_1 + [-5:\infty] \cdot \lambda_2 + [-3] \\
 \xrightarrow{\text{pointify}} & [0:3] \cdot \lambda_1 + [0:\infty] \cdot \lambda_2 + [4]\lambda_1 + [-5]\lambda_2 + [-3] \\
 \xrightarrow{\text{sweep}} & [0:6] \quad + [0:\infty] \quad + [4]\lambda_1 + [-5]\lambda_2 + [-3]
 \end{aligned}$$

So consider

$$h : \mathbf{S}_T \rightarrow \mathbb{R}, \quad h(\lambda_1, \lambda_2) := 4\lambda_1 - 5\lambda_2 - 3$$

after transforming back to \mathbf{D}_T with $\bar{\lambda} = \mathbf{A}^{-1} \cdot (\bar{\mathbf{x}} - \mathbf{B})$ to \mathbf{D}_T

$$\forall \bar{\mathbf{x}} \in \mathbf{D}_T \quad \mathbf{x}_1 \cdot \mathbf{x}_2 \geq \frac{5 \cdot \mathbf{x}_1 - 13 \cdot \mathbf{x}_2 - 5}{11}$$

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tangent space :

- standalone linearizer
- prototypical implementation in C++
- ca 400 SLOC: small user interface, plotting
- ca 400 SLOC: core routines (`tangent_space.*`)
- ca 1000 SLOC: helper routines, Taylor models (`tm4ts.*`)
- using rational numbers from GMP
- currently supported basic functions:
+ , - , * , / , sin , cos , exp , abs , sgn

Usage: From standard input read strings like:

- **function - * x0 x0 sin x1 end**

(define multivariate function $f(\mathbf{x}_0, \mathbf{x}_1) = \mathbf{x}_0^2 - \sin(\mathbf{x}_1)$ using prefix notation)

- **args -7/10 2/10 end**

(Specify argument vector \mathbf{c}_x , here ($\mathbf{c}_x = (0.7, 0.2)$)

- **conflict 3/10**

(Specify a 'conflict' \mathbf{c}_y , usually with $\mathbf{c}_y \neq f(\mathbf{c}_x)$)

- **bounds 0 100/100 -60/100 end**

(Define lists of coordinates for recycling,

for variable \mathbf{x}_0 try to use bounds $\frac{100}{100} = 1$ and $\frac{-60}{100} = -0.6$)

- **effort 20**

(Restrict the precision in the linearization process

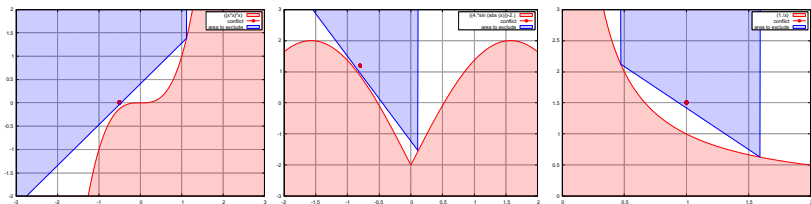
to radii larger than $2^{-20} \rightsquigarrow$ failure for $\mathbf{z} = f(\mathbf{x}_0, \mathbf{x}_1)$)

- **plot**

*(Give gnuplot commands and run them, plot area via **area**, 2d- or 3d-plot)*

Example inputs for tangent space

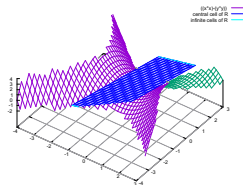
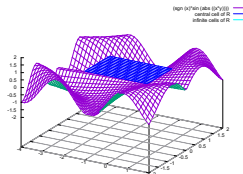
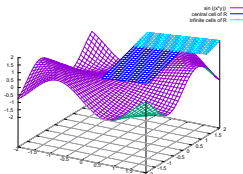
- predicates with two variables
- $f(\mathbf{x}_0) \diamond y$
- 2d-plot



- 1 function * * x0 x0 x0 end args -5/10 end conflict 0/1
area -3 3 -2 2 end show end plot
- 2 function - * 4 sin abs x0 2 end args -8/10 end conflict 12/10
ratio 10 1 area -2 2 -3 3 end plot end
- 3 function / 1 x0 end args 1 end conflict 3/2
area 0 2 0 3 end plot end

Example inputs for tangent space

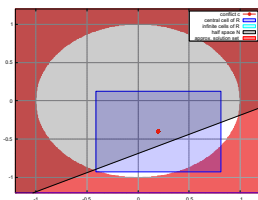
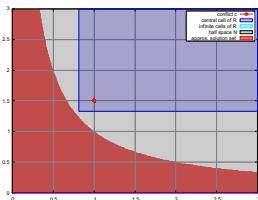
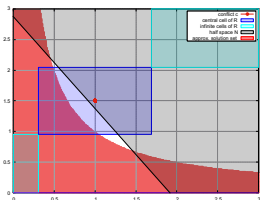
- predicates with three variables
- $f(\mathbf{x}_0, \mathbf{x}_1) \diamond \mathbf{y}$
- 3d-plot



- 1 function `sin * x0 x1 end args 1/2 1/5 end conflict 1/1`
area `-2 2 -2 2 end plot end`
- 2 function `* sgn x0 sin abs * x0 x1 end args 0 0 end conflict 1/1`
area `-2 2 -2 2 end plot end`
- 3 function `- * x0 x0 * x1 x1 end args 1/3 1/3 end conflict 2/1`
area `-4 3 -4 3 -2 4 end plot end`

Example inputs for tangent space,

- predicates with two variables, no \mathbf{c}_y
- $f(\mathbf{x}_0, \mathbf{x}_1) \diamond 0$
- 2d-plot



- 1 function - * x0 x1 1 end args 1 3/2 end
area 0 3 0 3 end plot end
- 2 function - * x0 x1 1 end args 1 3/2 end
interval area 0 3 0 3 end plot end
- 3 function - + * x0 x0 * x1 x1 1 end args 1/5 -2/5 end
area -1.2 1.2 -1.2 1.2 end plot end

- 1 Linearizing non-linearity
- 2 Notes on Interval arithmetic
- 3 Representing multivariate functions
- 4 Software demonstration
- 5 Outlook

Current state:

- currently: linear models with rational coefficients
- under construction: higher order models with computable coefficients

ToDo:

- implementing affine transformation
- optimization for cells
- integration in SMT solver
- composition, limits, ...

Thank you for your attention!

Questions?

Remarks?