

Interval Analysis

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Interval analysis

Interval arithmetic primer

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- Basic operations

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- Example

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- Parameter bounding

- Robust parameter bounding

- Sivia

- Sivia with contractors

- Example

Interval analysis

Provides efficient techniques to

- ▶ perform **guaranteed deterministic global** optimization,
- ▶ evaluate **all** solutions of a set of nonlinear equations
- ▶ compute **inner and outer** approximation of the set of vectors consistent with a set of inequalities
- ▶ ...

Interval analysis

Has lead to numerous applications

- ▶ Bounded-error parameter and state estimation of nonlinear systems
- ▶ Robust bounded-error parameter and state estimation
- ▶ Parameter estimation by global optimization
- ▶ Structural identifiability study
- ▶ Distributed estimation
- ▶ ...

Interval arithmetic primer

Introduced by Sunaga in Japan and by Moore in the USA.

Limited impact until beginning of the 90s

⇒ various reasons, among which implementation issues

Many books, code libraries, lists

<http://www.cs.utep.edu/interval-comp/main.html>

Ongoing standardization process IEEE P1788.

Interval of real numbers

Closed and *bounded* subset of \mathbb{R}

$$[x] = [\underline{x}, \bar{x}] = \{x \in \mathbb{R} | \underline{x} \leq x \leq \bar{x}\}.$$

It is a set \implies notions such as

$$=, \in, \subset, \cap$$

are well defined.

When considering \cup

$$[x] \cup [y] = [\min(\underline{x}, \underline{y}), \max(\bar{x}, \bar{y})].$$

Interval of real numbers

Other characteristics of an interval

Width

$$w([x]) = \bar{x} - \underline{x},$$

Midpoint

$$m([x]) = \frac{\underline{x} + \bar{x}}{2}.$$

Basic operations

Extended to intervals

$$\circ \in \{+, -, \times, /\}, [x] \circ [y] = \{x \circ y | x \in [x] \text{ and } y \in [y]\}.$$

More specifically

$$\left\{ \begin{array}{l} [x] + [y] = [\underline{x} + \underline{y}, \bar{x} + \bar{y}], \\ [x] - [y] = [\underline{x} - \bar{y}, \bar{x} - \underline{y}], \\ [x] \times [y] = [\min(\underline{x}.\underline{y}, \bar{x}.\underline{y}, \underline{x}.\bar{y}, \bar{x}.\bar{y}), \max(\underline{x}.\underline{y}, \bar{x}.\underline{y}, \underline{x}.\bar{y}, \bar{x}.\bar{y})], \\ [x] / [y] = [x] \times [1/\bar{y}, 1/\underline{y}], \text{ if } 0 \notin [y] \text{ and undefined else..} \end{array} \right.$$

Inclusion function

Range of a function over an interval

$$f([x]) = \{f(x) \mid x \in [x]\}$$

\Rightarrow difficult to obtain in general

\Rightarrow sometimes even not an interval

Inclusion function $[f](\cdot)$ of $f(\cdot)$ satisfies

$$\forall [x] \subset \mathbb{R}, f([x]) \subset [f]([x]).$$

Inclusion function is **minimal** if \subset may be replaced by $=$.

Convergent inclusion function

if $w([x]) \rightarrow 0$, then $w([f]([x])) \rightarrow 0$.

Inclusion function

Inclusion function **easy** to build for monotone functions

$$\begin{aligned}\sqrt{[x]} &= [\sqrt{\underline{x}}, \sqrt{\bar{x}}], \text{ if } \underline{x} \geq 0, \\ \exp([x]) &= [\exp(\underline{x}), \exp(\bar{x})], \\ \tan([x]) &= [\tan(\underline{x}), \tan(\bar{x})], \text{ if } [x] \subseteq [-\pi/2, \pi/2].\end{aligned}$$

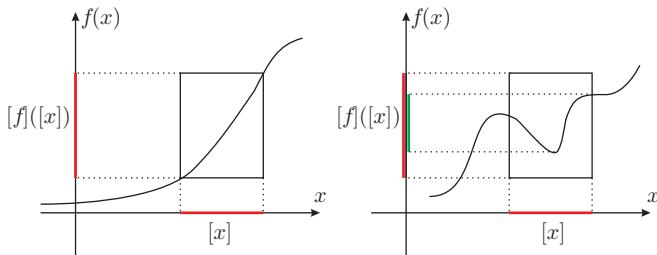
More complicated for other elementary functions

⇒ algorithm required for \sin, \cos, \dots

⇒ natural inclusion function

Inclusion function

Usually, an inclusion function is **not** minimal



\Rightarrow some overestimation of the range (**pessimism**).

Natural inclusion function

\Downarrow

Replace each real variable by its interval counterpart

$x \longrightarrow [x]$

Example

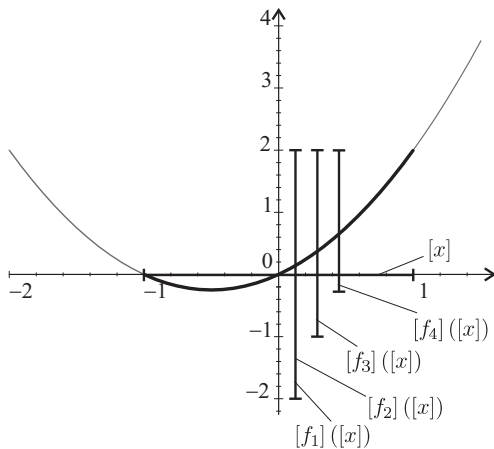
$$\begin{aligned}f_1(x) &= x(x+1), & f_3(x) &= x^2 + x, \\f_2(x) &= x \times x + x, & f_4(x) &= (x + \tfrac{1}{2})^2 - \tfrac{1}{4}.\end{aligned}$$

Results for $[x] = [-1, 1]$

$$\begin{aligned}[f_1]([x]) &= [x]([x] + 1) = [-2, 2], \\[f_2]([x]) &= [x] \times [x] + [x] = [-2, 2], \\[f_3]([x]) &= [x]^2 + [x] = [-1, 2], \\[f_4]([x]) &= ([x] + \tfrac{1}{2})^2 - \tfrac{1}{4} = [-\tfrac{1}{4}, 2].\end{aligned}$$

Only $[f_4](.)$ is minimal \iff **minimum number of occurrences** of the interval variable

Example



Centred form

For $f : \mathcal{D} \rightarrow \mathbb{R}$, differentiable over $[x] \subset \mathcal{D}$, one has $\forall x, m \in [x], \exists \xi \in [x]$ such that

$$f(x) = f(m) + (x - m) f'(\xi).$$

Then

$$f(x) \in f(m) + (x - m) f'([x]),$$

and

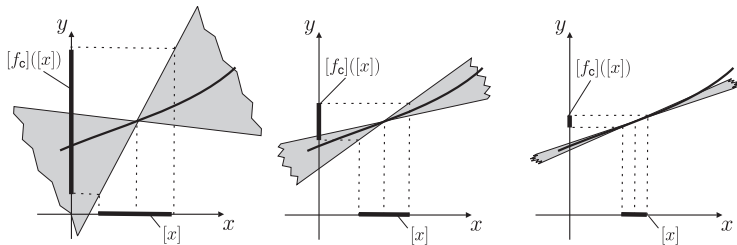
$$f([x]) \subseteq f(m) + ([x] - m) [f']([x]).$$

Centred form is the inclusion function defined by

$$[f]_c([x]) = f(m) + ([x] - m) [f']([x])$$

Centred form

Interpretation of the centred form



Example

Consider

$$f(x) = x^2 \exp(x) - x \exp(x^2).$$

Compare the natural inclusion function and the centred form

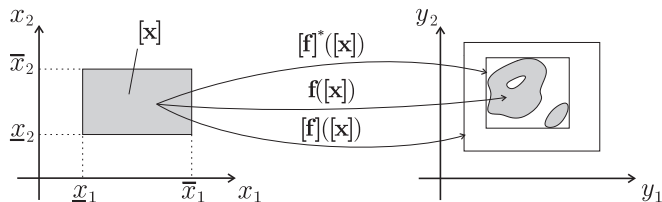
$[x]$	$f([x])$	$[f]([x])$	$[f]_c([x])$
$[0.5, 1.5]$	$[-4.148, 0]$	$[-13.82, 9.44]$	$[-25.07, 25.07]$
$[0.9, 1.1]$	$[-0.05380, 0]$	$[-1.697, 1.612]$	$[-0.5050, 0.5050]$
$[0.99, 1.01]$	$[-0.0004192, 0]$	$[-0.1636, 0.1628]$	$[-0.004656, 0.004656]$

Extension to vectors of intervals

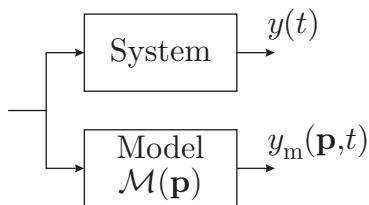
Vector of intervals or **box**

$$[\mathbf{x}] = [x_1] \times \cdots \times [x_n].$$

Vector inclusion function



Parameter estimation



\mathbf{y} : vector of experimental data

\mathbf{p} : vector of **unknown, constant** parameters

$\mathbf{y}_m(\mathbf{p})$: vector of model output

Parameter estimation :
Determination of $\hat{\mathbf{p}}$ from \mathbf{y} .

Problem formulation

1. Minimisation of a cost function, e.g.,

$$\hat{\mathbf{p}} = \arg \min_{\mathbf{p}} j(\mathbf{p}) = (\mathbf{y} - \mathbf{y}_m(\mathbf{p}))^T (\mathbf{y} - \mathbf{y}_m(\mathbf{p}))$$

- ▶ Local techniques : Gauss-Newton, Levenberg-Marquardt...
- ▶ Random search : simulated annealing, genetic algorithms...
- ▶ Global guaranteed techniques : Hansen's algorithm

2.

Parameter bounding

Experimental data : $y(t_i)$,

$t_i, i = 1 \dots, N$, known measurement times

$[\varepsilon_i] = [\underline{\varepsilon}_i, \bar{\varepsilon}_i], i = 1, \dots, N$, known **acceptable** errors

$\mathbf{p} \in \mathcal{P}_0$ deemed **acceptable** if for all $i = 1, \dots, N$,

$$\underline{\varepsilon}_i \leq y(t_i) - y_m(\mathbf{p}, t_i) \leq \bar{\varepsilon}_i.$$

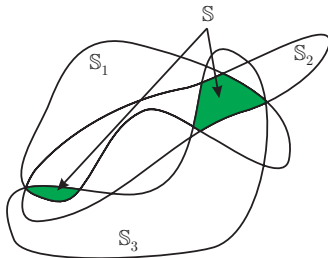
\implies Bounded-error parameter estimation :

Characterize $\mathbb{S} = \{\mathbf{p} \in \mathcal{P}_0 \mid y(t_i) - y_m(\mathbf{p}, t_i) \in [\underline{\varepsilon}_i, \bar{\varepsilon}_i], i = 1, \dots, N\}$

Parameter bounding

- ▶ When $y_m(\mathbf{p}, t_i)$ is linear in \mathbf{p}
 - ▶ exact description by polytopes
(Walter and Piet-Lahanier, 1989. . .)
 - ▶ outer approximation by ellipsoids, polytopes, ...
(Schweppe, 1973 ; Fogel and Huang, 1982. . .)
- ▶ When $y_m(\mathbf{p}, t_i)$ is non-linear in \mathbf{p}
 - ▶ outer approximation by polytopes, ellipsoids. . .
(Norton, 1987 ; Clément and Gentil, 1988 ; Cerone, 1991...)
 - ▶ approximate but guaranteed enclosure of \mathbb{S} by SIVIA
(Moore, 1992 ; Jaulin and Walter 1993)

Robust parameter bounding



$$\mathbb{S} = \bigcap_{\ell=1 \dots N} \mathbb{S}_\ell,$$

with

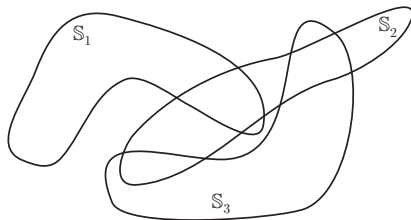
$$\mathbb{S}_\ell = \{\mathbf{p} \in \mathcal{P}_0 \mid y_\ell^m(\mathbf{p}) - y_\ell \in [\underline{\varepsilon}_\ell, \bar{\varepsilon}_\ell]\}$$

Interval analysis [2, 3], [1] allows to get

$$\underline{\mathbb{S}} \subset \mathbb{S} \subset \bar{\mathbb{S}}$$

No consistent \mathbf{p} is missed \implies **guaranteed set estimate.**

Robust parameter bounding

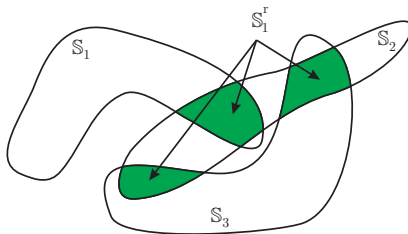


When the solution set is empty

$$\mathbb{S} = \bigcap_{\ell=1 \dots N} S_{\ell} = \emptyset.$$

Hypothesis on model or noise violated

Robust parameter bounding



Estimator robust against n outliers

$$\mathbb{S}_n^r = \bigcup_{1 \leq \ell_1 < \dots < \ell_n \leq N} \bigcap_{\ell \neq \ell_1, \dots, \ell_n} \mathbb{S}_\ell.$$

Intersection of $N - n$ sets among N

Interval analysis \Rightarrow **non-combinatorial** solution

$$\mathbb{S}_n^r = \left\{ \mathbf{p} \in \mathcal{P}_0 \mid \sum_{\ell=1}^N t_\ell(\mathbf{p}) \geq N - n \right\}$$

with

$$t_\ell(\mathbf{p}) = (y_\ell^m(\mathbf{p}) - y_\ell \in [\underline{\varepsilon}_\ell, \bar{\varepsilon}_\ell])$$

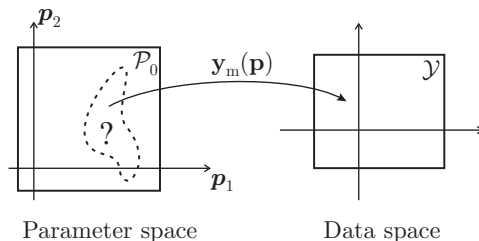
\mathbb{S}_n^r evaluated with a complexity of the order of that of \mathbb{S}

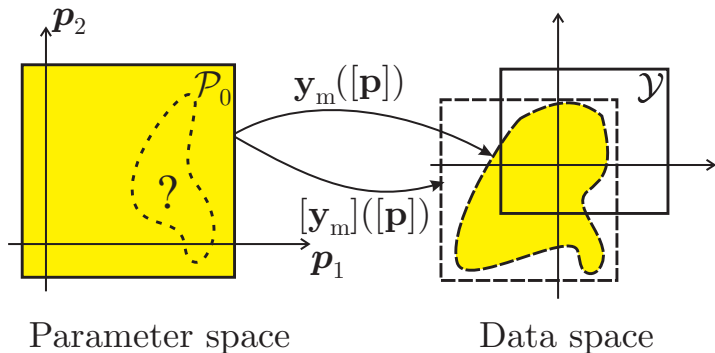
Set to be characterized

$$\begin{aligned}\mathbb{S} &= \{\mathbf{p} \in \mathcal{P}_0 \mid y(t_i) - y_m(\mathbf{p}, t_i) \in [\underline{\varepsilon}_i, \bar{\varepsilon}_i], i = 1, \dots, N\} \\ &= \{\mathbf{p} \in \mathcal{P}_0 \mid \mathbf{y}_m(\mathbf{p}) \subset \mathcal{Y}\},\end{aligned}$$

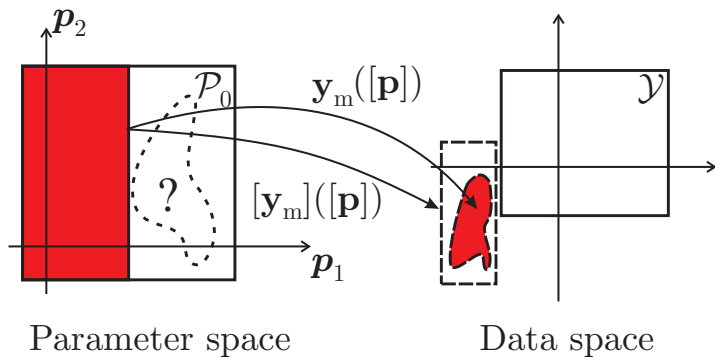
with

$$\mathcal{Y} = [y(t_1) - \bar{\varepsilon}_1, y(t_1) - \underline{\varepsilon}_1] \times \dots \times [y(t_N) - \bar{\varepsilon}_N, y(t_N) - \underline{\varepsilon}_N]$$

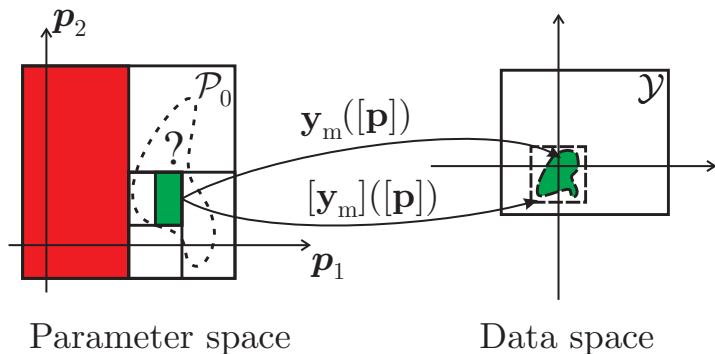




Yellow box is **undetermined**

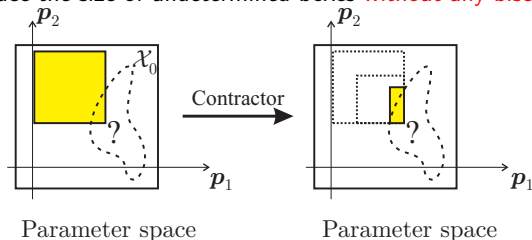


Red box **proven** to be outside \mathcal{S}



Green box **proven** to be included in \mathcal{S}

Reduce the size of undetermined boxes **without any bisection**



Contractors (Jaulin *et al*, 2001) based on

- ▶ interval constraint propagation (Walz)
- ▶ linear programming
- ▶ parallel linearization
- ▶ ...

Example of interval constraint propagation

$$y_m(\mathbf{p}) = p_1 \exp(-p_2),$$

$$p_1 \in [p_1]^0 = [-2, 2], \quad p_2 \in [p_2]^0 = [-2, 2].$$

One want to characterize the set

$$\mathbb{S} = \left\{ \mathbf{p} \in [p_1]^0 \times [p_2]^0 \mid \mathbf{y}_m(\mathbf{p}) \subset [1, 2] \right\}.$$

One may write that

$$p_1 \exp(-p_2) \in [1, 2],$$

thus

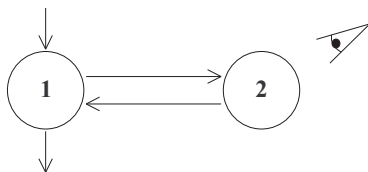
$$\begin{aligned} p_1 &\in [-2, 2] \cap \left(\frac{[1, 2]}{\exp(-[-2, 2])} \right) = [-2, 2] \cap [0.1353, 14.78] \\ &\in [0.1353, 2]. \end{aligned}$$

Similarly for p_2 , one has

$$\begin{aligned} p_2 &\in [-2, 2] \cap \left(-\ln \left(\frac{[1, 2]}{[0.1353, 2]} \right) \right) = [-2, 2] \cap [-2.6932, 0.6932] \\ &\in [-2, 0.6932] \end{aligned}$$

Example

Estimation of the parameters of a compartmental model



State equation

$$\begin{cases} x_1' = -(k_{01} + k_{21})x_1 + k_{12}x_2 \\ x_2' = k_{21}x_1 - k_{12}x_2 \end{cases} \quad \text{with} \quad \begin{cases} x_1(0) = 0 \\ x_2(0) = 0 \end{cases}$$

Observation equation

$$y(t_i) = x_2(t_i) + b(t_i), \quad i = 1, \dots, 16$$

Example

Model

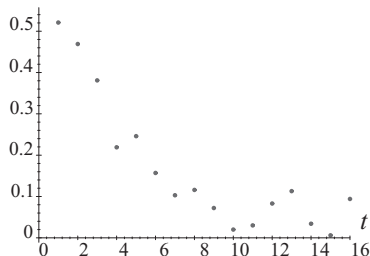
$$y_m(\mathbf{p}, t_i) = p_1 (\exp(p_2 t_i) - \exp(p_3 t_i)), \quad i = 1, \dots, 16,$$

where the macroparameters

$$\mathbf{p} = (p_1, p_2, p_3)^T$$

depends on the microparameters

$$(k_{01}, k_{12}, k_{21}).$$



Simulated
noisy experimental data

Example

Macroparameter estimation with

$$\underline{\varepsilon}_i = -0.09, \bar{\varepsilon}_i = 0.09, i = 1, \dots, 16$$

Results

	SIVIA	SIVIA + ICP	ICP only
Comp. time (s)	8	6.2	0.63
Bounding box	$[0.49, 1.06]$ $[-0.293, -0.141]$ $[-5, -1.054]$	$[0.49, 1.06]$ $[-0.293, -0.141]$ $[-5, -1.054]$	$[0.52, 0.98]$ $[-0.282, -0.156]$ $[-5, -1.167]$

Conclusions

- ▶ Interval techniques provide guaranteed solution to bounded-error parameter estimation
- ▶ Robust estimation possible
- ▶ ICP or SIVIA + ICP allows more unknown parameters than SIVIA

Also possible

- ▶ Bounded-error state estimation
- ▶ Parameter estimation via deterministic global optimization

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