

# The Optimization Test Environment

Martin Fuchs

in collaboration with Ferenc Domes, Hermann Schichl,  
University of Vienna

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## Motivation

Testing is crucial for users who seek to

- ① compare solvers on suitable test sets  
→ solver benchmarks with large test sets
- ② compare solvers on single problems  
→ find a good solution of a specific problem
- ③ adjust solver parameters  
→ improve the configuration of a particular solver

The Test Environment automatizes many of these (possibly little exciting) procedures for global optimization solvers.



## Study background

Comparisons of global optimization solvers:

- one extensive study in 2005, based on the COCONUT Environment benchmark
- typically for black box problems only
- many interfaces of test environments are not very user friendly

! Generally, much effort has to be spent to conduct a solver comparison



## Basic testing procedure

- ① organize (possibly large) test libraries
- ② solve selected problems with selected solvers
- ③ analyze, check and compare the results
- ④ create summary tables (e.g., in  $\text{\LaTeX}$ )



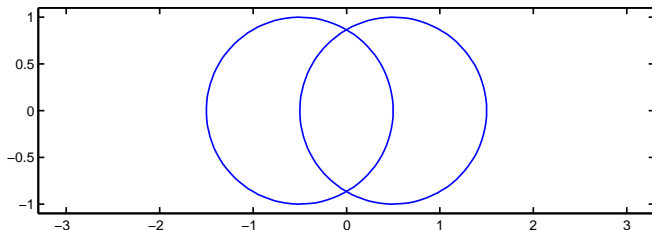
## Notation for optimization problems

$$\left. \begin{array}{l} \min f(x) \\ \text{s.t. } x \in \mathbf{x}, \\ F(x) \in \mathbf{F}, \\ x_i \in \mathbb{Z} \text{ for } i \in I. \end{array} \right\} (1)$$

- $\mathbf{x} = [\underline{x}, \bar{x}] = \{x \in \mathbb{R}^n \mid \underline{x} \leq x \leq \bar{x}\}$  is a **box** in  $\mathbb{R}^n$
- $f : \mathbf{x} \rightarrow \mathbb{R}$  is the **objective function**
- $F : \mathbf{x} \rightarrow \mathbb{R}^m$  is a vector of **constraint functions**
- $\mathbf{F}$  is a box in  $\mathbb{R}^m$
- $I \subseteq \{1, \dots, n\}$  defines the **integer components** of  $x$

## Test problem examples

$$\left. \begin{array}{l} \min_x x_2 \\ \text{s.t. } (x_1 - 0.5)^2 + x_2^2 = 1, \\ (x_1 + 0.5)^2 + x_2^2 = 1, \\ x_1 \in [-3, 3], x_2 \in [-3, 3]. \end{array} \right\} (2)$$



$$\rightarrow \hat{x} = (0, -\sqrt{3}/2)^T \approx (0, -0.8660)^T$$



## AMPL formulation

- var x1  $\geq -3$ ,  $\leq 3$ ;

- var x2  $\geq -3$ ,  $\leq 3$ ;

minimize obj: x2;

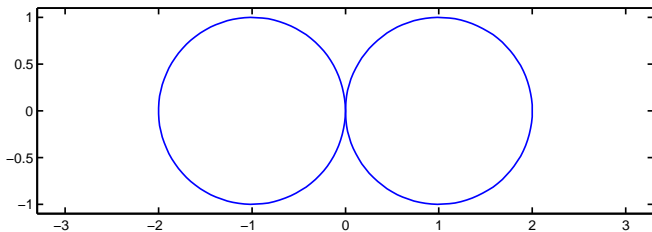
s.t. c1:  $(x1-0.5)^2+x2^2=1$ ;

s.t. c2:  $(x1+0.5)^2+x2^2=1$ ;

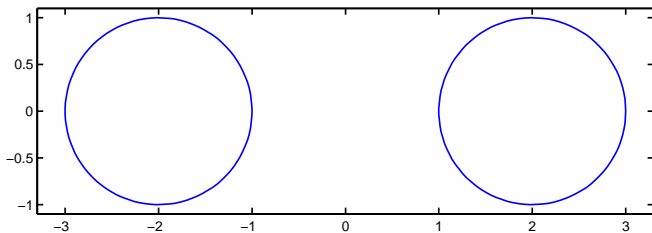
- our problem representations are DAGs (COCONUT Environment)
- converters to DAGs will be available online for popular modeling languages (e.g., AMPL, GAMS)



## Test problem examples ctd.



→  $x = (0, 0)^T$  is the only feasible point



→ no feasible solution





# Interactive library management

[demo]



## Solver interface

- ① solver selection
- ② solver configuration  
→ setup paths and scripts
- ③ test problem subset selection  
→ criteria editor
- ④ solve command  
→ solve selected problems with selected solvers

→ [demo]



## Analyze

- translate solver results to a unified format

→ .res files

modelstatus	solver model status
x(i)	solver output for $\hat{x}_i, i = 1, \dots, n$
obj	solver output for $f(\hat{x})$
infeas	feasibility distance provided by the solver
time	used CPU time to solve the problem
splits	number of splits made



## Solution check

- let  $C := \{x \in \mathbf{x} \mid x_I \in \mathbb{Z}, F(x) \in \mathbf{F}\}$
- compute a feasibility distance  $d_{\text{feas}}$
- intuitively:  $\min_{y \in C} \|x - y\|_p$
- computationally: componentwise feasibility
- account for scaling issues



## Solution check ctd.

- parameters  $\varepsilon$  ( $:= 10^{-6}$ ),  $\kappa$  ( $:= 1$ ),  $\alpha$  ( $:= 10^{-6}$ )
- solver result  $x_s, f_s$
- $\mathbf{x}_s = x_s \pm \varepsilon \max(\|x_s\|_\infty, \kappa)$
- objective violation  

$$v_o := \langle f(x_s) + f'(\mathbf{x}_s)(\mathbf{x}_s - x_s) - f_s \rangle$$
- box constraint violations  

$$v_b := \langle \mathbf{x}_s - \mathbf{x} \rangle$$
- constraint violations  

$$v_c := \langle F(x_s) + F'(\mathbf{x}_s)(\mathbf{x}_s - x_s) - \mathbf{F} \rangle$$
- violation  $v = \|(v_o, v_b, v_c)^T\|_p$  (we use  $p = \infty$ )
- if  $v \leq \alpha \Rightarrow x_s$  **feasible**

## Comparison of results

- $J = \{x_1, \dots, x_N\}$ :  
solver results that passed the check
- **global solution**  $x_{\text{opt}}$ :  
 $x_{\text{opt}} \in J$  and  $f(x_{\text{opt}}) \leq f(x_j)$  for all  $x_j \in J$
- $\tilde{x} \in J$ ,  $\tilde{x} \neq x_{\text{opt}}$  is also considered global if  $f(\tilde{x})$   
is sufficiently close to  $f(x_{\text{opt}})$
- otherwise  $\tilde{x}$  is a **local solution**



## Comparison of results ctd.

- **best point:**

$$x_{\text{best}} = \begin{cases} x_{\text{opt}} & \text{if } J \neq \emptyset, \\ \arg \min_{x \in J} d_{\text{feas}}(x) & \text{if } J = \emptyset. \end{cases}$$

- if the best point was found by a default local solver  $\rightarrow$  **easy location**
- otherwise  $\rightarrow$  **hard location**



## Performance assessment

Count for each solver

- the number of global solutions found
- the number of wrong and correct solver claims

Further performance measures for rigorous or black box solvers can be implemented in the future.





## Summary tables

- automated creation of summary tables of the results in  $\text{\LaTeX}$
- just copy paste them to result sections of scientific publications
- [demo]



## Benchmark

### COCONUT Environment benchmark:

- three libraries of global constrained optimization and constraint satisfaction problems
- over 1000 test problems in up to 20000 variables
- time limits depending on the problem size:

size	$n$	timeout (s)
1	1-9	180
2	10-99	900
3	100-999	1800
4	$\geq 1000$	.

We will collect solver benchmark results online.



## Preliminary results

BARON summary statistics									
lib	all	acc	wr	G+	G!	!!	G?	L?	I?
Lib1	254	218	7	197	94	1	7	0	0
Lib2	599	532	21	440	206	4	19	1	1
Lib3	271	247	21	214	213	8	19	0	2

KNITRO summary statistics									
lib	all	acc	wr	G+	G!	!!	G?	L?	I?
Lib1	254	232	38	141	0	0	0	38	0
Lib2	599	587	111	357	0	0	0	111	0
Lib3	271	265	44	147	0	0	0	44	0

## Conclusions

### Easy-to-use management of

- test libraries and problem subset selection
- solver configurations

### Automatized

- solution check and comparison
- $\text{\LaTeX}$  code creation



## Final remarks

The preliminary benchmark results show

- global optimization is superior to local in most test cases
- BARON performs well as a global solver
- visit my website for updates

<http://www.martin-fuchs.net>