

# Verified Linear Programming

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# Outline

- 1 Introduction
- 2 Rigorous Bounds
- 3 Numerical experience
  - Netlib
  - Comparison
- 4 Software
- 5 Summary

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# Why linear programming?

- most basic form of optimization
- wide range of applications:  
oil refinery problems, flap settings on aircraft, industrial production and allocation, image restoration, linearization, linear relaxations in global optimization, ...

## Lovasz (1980)

*If one would take statistics about which mathematical problem is using up most of the computer time in the world, then (not including database handling problems like sorting and searching) the answer would probably be linear programming.*

## And why verified?

- Ordóñez and Freund (2003):  
71% of Netlib lp problems are ill-posed

### Ben-Tal and Nemirovski (2000)

*In real-world applications of Linear Programming one cannot ignore the possibility that a small uncertainty in the data (intrinsic for most real-world LP programs) can make the usual optimal solution of the problem completely meaningless from a practical viewpoint.*

# A representation of a linear program

## Definition (Linear Program)

Find the optimal value  $f^*$  of a linear objective function  $c^T x$  subject to

- linear constraints  $Ax \leq a$ ,  $Bx = b$  and
  - simple bounds  $\underline{x} \leq x \leq \bar{x}$ .
- 
- Set of feasible points  $F$  satisfying constraints and simple bounds
  - Can be represented by the tuple  $P := (c, A, a, B, b)$  and  $\underline{x}, \bar{x}$
  - Simple bounds **may be infinite**

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# A lower bound for the optimal value (1)

## Theorem (Jansson (2004))

Given a linear program  $P$  and  $\underline{x}, \bar{x}$ . If  $\mathbf{y} \leq 0, \mathbf{z}$  satisfy

- $\exists \mathbf{y} \in \mathbf{y}, \mathbf{z} \in \mathbf{z} : c_j - (A_{:j})^T \mathbf{y} - (B_{:j})^T \mathbf{z} = 0$  for free  $x_j$ , and
- the defects

$$\mathbf{d}_j := c_j - (A_{:j})^T \mathbf{y} - (B_{:j})^T \mathbf{z} \begin{cases} \leq 0 & \text{for } -\infty < x_j \leq \bar{x}_j \\ \geq 0 & \text{for } \underline{x}_j \leq x_j < \infty \end{cases}$$

then a lower bound for the optimal value is

$$\underline{f}^* := \inf \left\{ \mathbf{a}^T \mathbf{y} + \mathbf{b}^T \mathbf{z} + \sum_{\underline{x}_j \neq -\infty} \underline{x}_j \mathbf{d}_j^+ + \sum_{\bar{x}_j \neq \infty} \bar{x}_j \mathbf{d}_j^- \right\}.$$



## A lower bound for the optimal value (2)

- $O(n^2)$  operations for finite simple bounds, coincides with Neumaier and Shcherbina's (2004)
- Conditions not met (infinite simple bounds)  
⇒ iterate with perturbed dual constraints
- Works for interval problems **P**

# An upper bound for the optimal value

- Based on idea from Krawczyk (1975) later used and modified  
Jansson (1988), Hansen and Walster (1991),  
Kearfott (1994)
- Enclose primal interior point
- Verifies existence of primal feasible solutions
- Works for interval problems **P**

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# A real world application test set

Netlib comprises  $\sim 100$  problems

- First added in 1988, latest in 1996
- 32 to 15695 variables, 27 to 16675 constraints (medium size)
- Established test set for lp algorithms
- Ordóñez and Freund (2003): 71% ill-conditioned

# Condition and the distance to infeasibility

## Definition (Distance to Infeasibility $\rho$ )

Norm of the smallest perturbation of  $P$  that results in an empty feasible set.

$\rho_P$  distance to **primal** infeasibility

$\rho_D$  distance to **dual** infeasibility

## Definition (Condition)

Scale invariant reciprocal of the minimal distance to infeasibility.

$$C(d) := \frac{\|d\|}{\min\{\rho_P, \rho_D\}}$$

# Some examples of what can happen

	$\rho_D$	$\rho_P$	$\underline{f}^*$	$\bar{f}^*$	$\mu$
scagr25	0.034646	0.021077	✓	✓	$3.7821e-8$
bnl1	0.106400	0	✓		$7.2244e-8$
stair	0	0.000580		✓	$5.4796e-9$
80bau3b	0	0	✓	✓	$5.6653e-8$
scsd8	1.000000	0.268363	✓		$6.3831e-8$

	m	p	n	nz
scagr25	171	300	500	1554
bnl1	411	232	1175	5121
stair	147	209	467	3856
80bau3b	2262	0	9799	21002
scsd8	0	397	2750	8584

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# Overview of the results

In total 89 problems, 86 with infinite simple bounds

- 35 finite upper bounds for the optimal value

$$\text{med}(\mu(\bar{f}^*, f^*)) = 8.0e - 9 \quad \text{med}(t_{\bar{f}^*} / t_{f^*}) = 5.3$$

- 76 finite lower bounds for the optimal value

$$\text{med}(\mu(\underline{f}^*, f^*)) = 2.2e - 8 \quad \text{med}(t_{\underline{f}^*} / t_{f^*}) = 0.5$$

# Software to solve LP rigorously

- iCOs, RealPaver
  - constraint satisfaction codes
  - no optimization  $\Rightarrow$  lower bound from non-satisfiable problems
  - iCOs: constraint programming, interval analysis
  - RealPaver: branch-and-prune, interval Newton method

# Software to solve LP rigorously

- iCOs, RealPaver
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  - no optimization  $\Rightarrow$  lower bound from non-satisfiable problems
  - iCOs: constraint programming, interval analysis
  - RealPaver: branch-and-prune, interval Newton method
- perPlex
  - verification tool using rational arithmetic
  - checks feasibility and optimality of a (approximate) basis
  - requires rational solution
  - cannot handle interval data
  - proof of concept (cannot compute solution, proceed from suboptimal basis)

# More software to solve LP rigorously

- GlobSol, Numerica, COSY
  - global optimization
  - GlobSol: interval branch-and-bound, automatic differentiation, constraint propagation, interval Newton, ...
  - Numerica: interval methods (Hansen–Sengupta), constraint satisfaction
  - COSY: branch-and-bound, Taylor model arithmetic

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  - Numerica: interval methods (Hansen–Sengupta), constraint satisfaction
- Lurupa
  - verified linear programming in the presence of uncertainty

# A proper test set

- 99 random problems after Rosen and Suzuki
  - choose solution  $\Rightarrow$  generate problem from KKT
  - non-degenerate
  - exactly known, integral solution
  - 3 instances per dimension set (variables, inequalities, equations)
- 19 real-world problems from Netlib and Meszaros's collection
  - less than 50 variables
  - large finite search region
- 1 hour timeout

## Results for random problems

	solved	$t_{max}$
5 variables (total 36)	iCOs	36 0.091s
	RealPaver	34 1378.100s
	perPlex	36 0.028s
	GlobSol	36 548.582s
	Lurupa	36 < 0.010s

Numerica solved 5 inequalities and 5 variables in 326.7s



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	solved	$t_{max}$
10 variables (total 36)	iCOs	36 0.254s
	RealPaver	0
	perPlex	36 0.036s
	GlobSol	3 666.963s
	Lurupa	36 < 0.010s

## Results for larger random problems

	solved	$t_{max}$
25 variables (total 18)		
iCOs	5	0.273s
perPlex	18	0.077s
Lurupa	18	< 0.010s

### iCOs

- solved problems with 15 or 25 equations
- ran **out of memory** for the remaining problems

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### iCOs

- solved problems with 15 or 25 equations
- ran out of memory for the remaining problems

	solved	$t_{max}$
500 variables (total 9)	perPlex	0
	Lurupa	9 29.2s

## Results for real-world problems

	solved	$t_{max}$
feasible problems (total 15)	iCOs	7    0.401s
	RealPaver	12    207.500s
	perPlex	15    0.021s
	GlobSol	6    4.564s
	Lurupa	15    < 0.010s

iCOs **wrongly** claims 3 problems to be infeasible

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iCOs wrongly claims 3 problems to be infeasible

	terminated	$t_{max}$
infeasible problems (total 4)	iCOs	4    1.190s
	RealPaver	4    < 0.010s
	perPlex	0
	GlobSol	0
	Lurupa	4    0.040s

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Lurupa returns some wide bounds ( $[0, \infty]$ ,  $[-\infty, \infty]$ )

# Outline

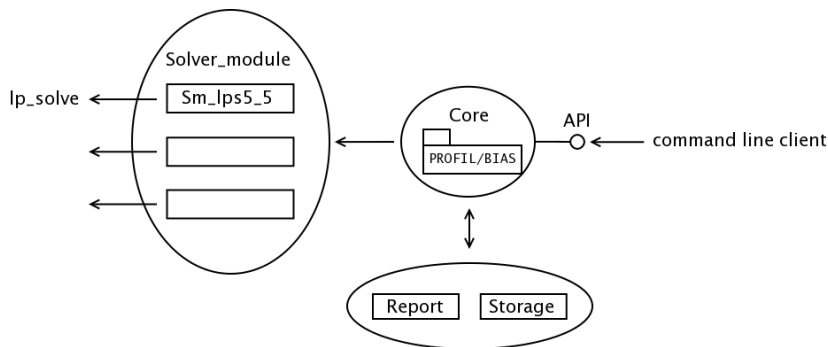
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# Lurupa's design goals

- Rigorous bounds for the optimal value
- Rigorous enclosure of near optimal, feasible points
- Use unmodified solvers plus postprocessing
- Easy to use
- Easily adoptable to different solvers
- Standalone and library version

# Lurupa's internals

- Implemented in ANSI C++
- Builds on interval library PROFIL/BIAS by Knüppel (1994)



# Usage of Lurupa

## Example (command line)

```
>lurupa -sm /path/to/module \  
        -lp /path/to/model -lb -ub
```

### Meaning of parameters

- sm specify solver module
- lp specify lp model file
- lb compute lower bound
- ub compute upper bound

# Usage of Lurupa

## Example (API)

```
Lurupa l;  
l.set_solver_module("path/to/module");  
  
FILE *in = fopen("path/to/model", "r");  
Lp lp = l.read_lp(in, 0);  
  
double optimal, bound, iterations;  
l.solve_lp(lp, optimal);  
l.lower_bound(lp, bound, iterations);  
l.upper_bound(lp, bound, iterations);
```

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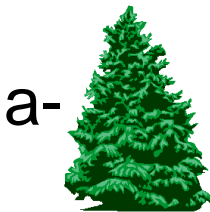
- Lurupa implements rigorous bounds and solves medium-scale problems in reasonable time
- General optimization software can verify solutions up to  $\sim 10 - 20$  variables
- Larger dimensions require to exploit special structure



# Future work

- Verified condition measures
- Sparse structures (in PROFIL)
- New solver modules
- Certificates of infeasibility, unboundedness
- Verified preprocessing as suggested by Fourer and Gay (1993)

# Thank you for your



# Dimensions for real-world problems

## Meszaros's misc problems

	m	p	n
kleemin3	3	0	3
kleemin4	4	0	4
kleemin5	5	0	5
kleemin6	6	0	6
kleemin7	7	0	7
kleemin8	8	0	8
farm	5	2	12
p0033	15	0	33
refine	29	0	33
p0040	23	0	40
problem	0	12	46

## Meszaros's infeas problems

	m	p	n
itest2	9	0	4
galenet	6	2	8
itest6	9	2	8
bgprtr	6	14	35

## Netlib problems

	m	p	n
afiro	19	8	32
kb2	27	16	41
sc50a	30	20	48
sc50b	30	20	48

# Detailed results for real-world problems

Meszaros's misc problems

	iCOs		RealPaver		perPlex	GlobSol	Lurupa
	t	$\Delta$	t	$\Delta$			
kleemin3	0.004	1e-12	0	1e-12	0.011	0.032	< 0.010
kleemin4	0.007	1e-12	0.030	1e-12	0.012	0.089	< 0.010
kleemin5	0.009	1e-12	0.210	1e-12	0.014	0.240	< 0.010
kleemin6	?		3.030	1e-12	0.012	0.639	< 0.010
kleemin7	?		207.500	1e-12	0.011	1.675	< 0.010
kleemin8	?		10.380	1e-4	0.012	4.564	< 0.010
farm	0.009	1e-12	1.480	1e-8	0.012	t	< 0.010
p0033	oom		t	1e-1	0.017	t	< 0.010
refine	oom		0	1e-1	0.021	e/t	< 0.010
p0040	oom		t	1e-1	0.014	t	< 0.010
problem	0.019	1e-12	0	1e-12	0.015	t	< 0.010

# More detailed results for real-world problems

## Meszaros's infeas problems

	iCOs	RealPaver	perPlex	GlobSol	Lurupa
itest2	0.947	0	lpinf	e/t	< 0.010s [0, ∞]
galenet	0	0	lpinf	e	< 0.010s [0, ∞]
itest6	1.190	0	lpinf	e/t	< 0.010s [1.9e6, ∞]
bgprtr	0.025	0	lpinf	t	0.040s [-∞, +∞]

## Netlib problems

	iCOs		RealPaver		perPlex	GlobSol	Lurupa
	t	$\Delta$	t	$\Delta$			
afiro	oom		55.680	1e-12	0.012	t	< 0.01
kb2	oom		t	1e-1	0.016	t	< 0.01
sc50a	0.401	1e-12	157.700	1e-2	0.016	t	0.01
sc50b	0.342	1e-12	0.740	1e-1	0.016	t	< 0.01