



www.localsolver.com

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Focused on business needs

A solver aligned with enterprise needs

- Provides high-quality solutions in seconds
- Scalable: tackles problems with millions of decisions
- Proves optimality when possible (best effort)

A solver aligned with practitioner needs

- “Model & Run”
 - Simple mathematical modeling formalism
 - Direct resolution: no need of complex tuning
- Simple and transparent pricing

Free for academics



Main features

New-generation solver

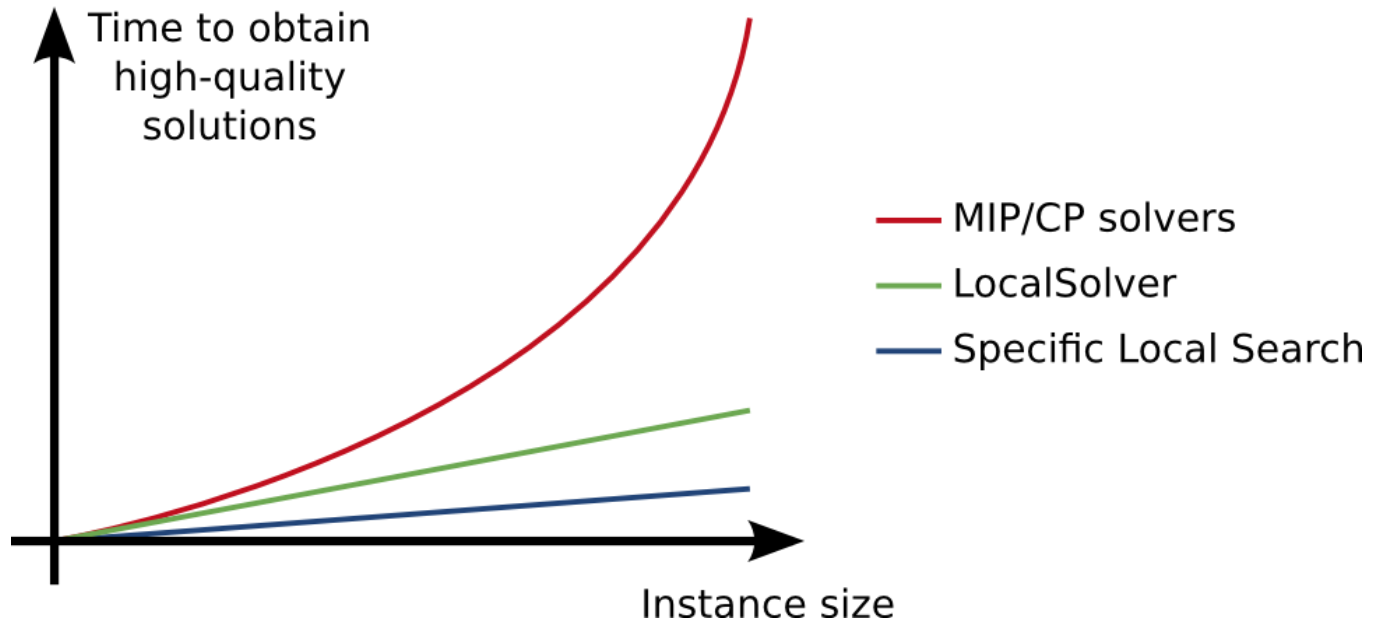
- **Computing good-quality solutions by local search**
- Computing lower bounds **separately** (inference, relaxation, cuts)

High-end software

- An innovative modeling language for fast prototyping
- Lightweight object-oriented APIs: a few classes only
- Reliable and robust: quality through drastic continuous integration
- Fully portable: Windows, Linux, Mac OS (x86, x64)
- Reactive support, realized by developers themselves (even for academics)



Why local search?



Technology

Autonomous local search

- Generic moves based on decisions/constraints hypergraph
- Incremental evaluation: **million moves per minute**
- Adaptive simulated annealing through learning
- Multithreaded search: ready for many-core world

Efficient C++ implementation

- Preprocessing: model reduction & reformulation
- Low-level cache-aware code optimization
- Highly-optimized memory management



Knapsack

8 items to pack in a sack: maximize the total value of items while not exceeding a total weight of 102 kg

```
function model() {  
  // 0-1 decisions  
  x_0 <- bool(); x_1 <- bool(); x_2 <- bool(); x_3 <- bool();  
  x_4 <- bool(); x_5 <- bool(); x_6 <- bool(); x_7 <- bool();  
  
  // weight constraint  
  knapsackWeight <- 10*x_0 + 60*x_1 + 30*x_2 + 40*x_3 + 30*x_4 + 20*x_5 + 20*x_6 + 2*x_7;  
  constraint knapsackWeight <= 102;  
  
  // maximize value  
  knapsackValue <- 1*x_0 + 10*x_1 + 15*x_2 + 40*x_3 + 60*x_4 + 90*x_5 + 100*x_6 + 15*x_7;  
  maximize knapsackValue;  
}
```

Binary decision variables

Integer intermediate variables

The user writes the model: nothing else to do!
declarative approach = model & run

Multiobjective knapsack

```
function model() {  
  // 0-1 decisions  
  x[0..7] <- bool();  
  
  // weight constraint  
  knapsackWeight <- 10*x[0]+ 60*x[1]+ 30*x[2]+ 40*x[3]+ 30*x[4]+ 20*x[5]+ 20*x[6]+ 2*x[7];  
  constraint knapsackWeight <= 102;  
  
  // maximize value  
  knapsackValue <- 1*x[0]+ 10*x[1]+ 15*x[2]+ 40*x[3]+ 60*x[4]+ 90*x[5]+ 100*x[6]+ 15*x[7];  
  maximize knapsackValue;  
  
  // secondary objective: minimize product of minimum and maximum values  
  knapsackMinValue <- min[i in 0..7](x[i] ? values[i] : 1000);  
  knapsackMaxValue <- max[i in 0..7](x[i] ? values[i] : 0);  
  knapsackProduct <- knapsackMinValue * knapsackMaxValue;  
  minimize knapsackProduct;  
}
```

Nonlinear operators: prod, min, max,
and, or, if-then-else, ...

Lexicographic objectives



Mathematical operators

Arithmetic		Logical	Relational	Hybrid
sum	prod	not	==	if
min	max	and	!=	array + at
div	mod	or	<=	
abs	sqrt	xor	>=	
			<	
			>	



Modeling APIs

```
function model() {  
  // 0-1 decision  
  x[1..nbltms]  
  
  // weight constraint  
  knapsackWeight  
  constraint kn  
  
  // maximize  
  knapsackValue  
  maximize kn  
}
```

```
#include "localsolver.h"  
using namespace localsolver;
```

C++

```
import localsolver.*;
```

Java

```
public class Toy {
```

C#

```
using System;  
using localsolver;
```

```
public class Toy
```

```
{  
  static void Main()
```

```
{  
  int[] weights = {10, 60, 30, 40, 30, 20, 20, 2};  
  int[] values = {1, 10, 15, 40, 60, 90, 100, 15};
```

```
  LocalSolver localsolver = new LocalSolver();  
  LSModel model = localsolver.GetModel();
```

```
  // 0-1 decisions  
  LSExpression[] x = new LSExpression[8];  
  for (int i = 0; i < 8; i++)  
    x[i] = model.CreateExpression(LSOperator.Bool);
```

createExpression

```
  // knapsackWeight <- 10*x0 + 60*x1 + 30*x2 + 40*x3 + 30*x4 + 20*x5 + 20*x6 + 2*x7;  
  LSExpression knapsackWeight = model.CreateExpression(LSOperator.Sum);
```

```
  for (int i = 0; i < 8; i++)  
    knapsackWeight.AddOperand(model.CreateExpression(LSOperator.Prod, weights[i], x[i]));
```

addOperand

```
  // knapsackWeight <= 102;  
  model.AddConstraint(model.CreateExpression(LSOperator.Leq,
```

```
  // knapsackValue <- 1*x0 + 10*x1 + 15*x2 + 40*x3 + 60*x4 + 90*x5 + 100*x6 + 15*x7;  
  LSExpression knapsackValue = model.CreateExpression(LSOperator.Sum);
```

```
  for (int i = 0; i < 8; i++)  
    knapsackValue.AddOperand(model.CreateExpression(LSOperator.Prod, values[i], x[i]));
```

```
  // maximize knapsackValue;  
  model.AddObjective(knapsackValue, LSObjectiveDirection.Maximize);
```

```
  // close the model before solving it  
  model.Close();
```

```
  LSPhase phase = localsolver.CreatePhase();  
  phase.SetTimeLimit(1);  
  localsolver.Solve();
```

```
}
```



Car sequencing in Renault's plants



Scheduling cars along painting and assembly lines

- Classical car sequencing = space car options along the line
- No more than K consecutive cars with the same color
- Minimize the number of paint color changes as secondary objective



Large-instances to tackle

- 1300 cars to sequence → 400 000 binary decisions
- MIP or CP solvers unable to find feasible solutions after hours
- LocalSolver provides much better solutions than Renault in seconds



2012 ROADEF/EURO Challenge

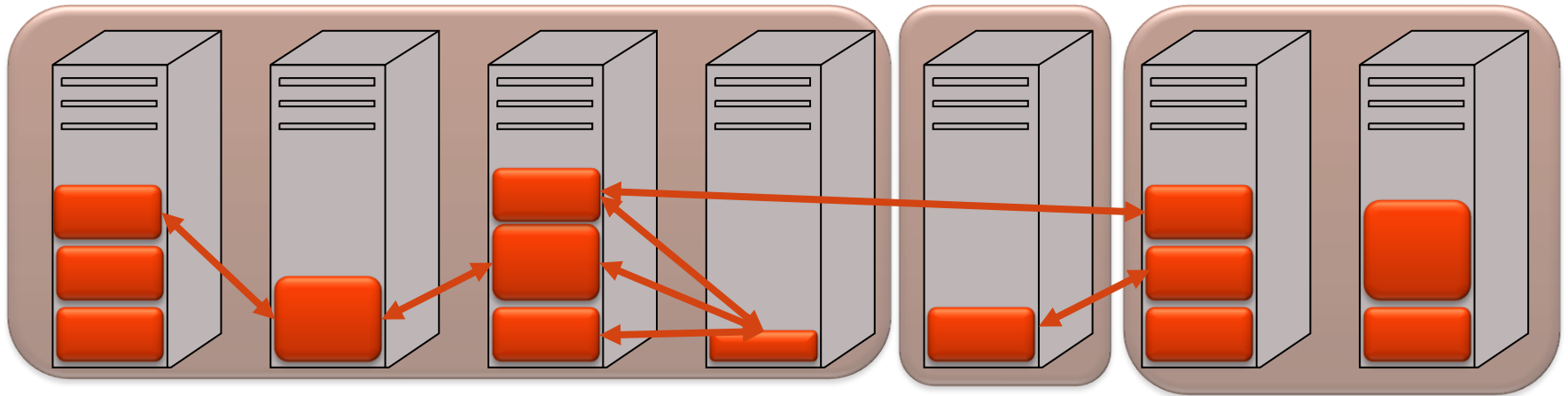


EURO



Google

Reassignment of processes to machines, with different kinds of constraints (mutual exclusion, resources, etc.)



More than 100 000 binary decisions

LSP model with 200 lines, written in 1 day of work

LocalSolver qualified for final round (ranked 25/80)

When using LocalSolver?

- MIP solvers find no (quality) solution
- MIP solvers find quality solutions but too slowly
- Writing MIP models is complicated due to nonlinearities
- CP seems to be a better choice than MIP

LocalSolver is suited for:

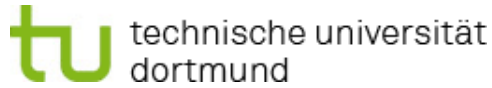
- Nonlinear assignment: car sequencing, frequency assignment
- Packing & Covering: media planning, machine scheduling, graph partitioning
- Facility location, logistic clustering, telecom network optimization
- Workforce scheduling, group planning, nurse rostering



Customers & Partners



Academic Users



東北大学
TOHOKU UNIVERSITY



UCC
Coláiste na hOllscoile Corcaigh, Éire
University College Cork, Ireland



UNSW
THE UNIVERSITY OF NEW SOUTH WALES
SYDNEY · CANBERRA · AUSTRALIA



UPPSALA
UNIVERSITET



POLITECNICO
DI MILANO

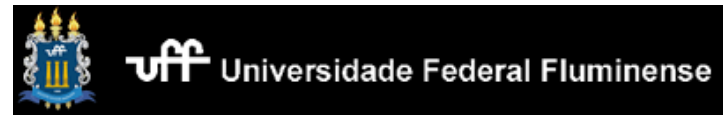


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Fast-growing community!

TARBIAT MODARES UNIVERSITY



IFSTTAR



LocalSolver

Roadmap

LocalSolver 3.0 : October 2012

- Floating-point coefficients
- New math operators: log, exp, pow, cos, sin, tan
- Major performance improvements: new local-search moves
- Improved preprocessing (model reduction & reformulation)

LocalSolver 4.0 : March 2013

- Binary + **continuous** decisions
 - Local-search moves on continuous decisions
 - Better capabilities for proving optimality or infeasibility
- **Large-scale mixed-variable nonlinear programming (MINLP)**



For more details



T. Benoist, B. Estellon, F. Gardi, R. Megel, K. Nouioua.
LocalSolver 1.x: a black-box local-search solver for 0-1 programming.
4OR, A Quarterly Journal of Operations Research 9(3), pp. 299-316.

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