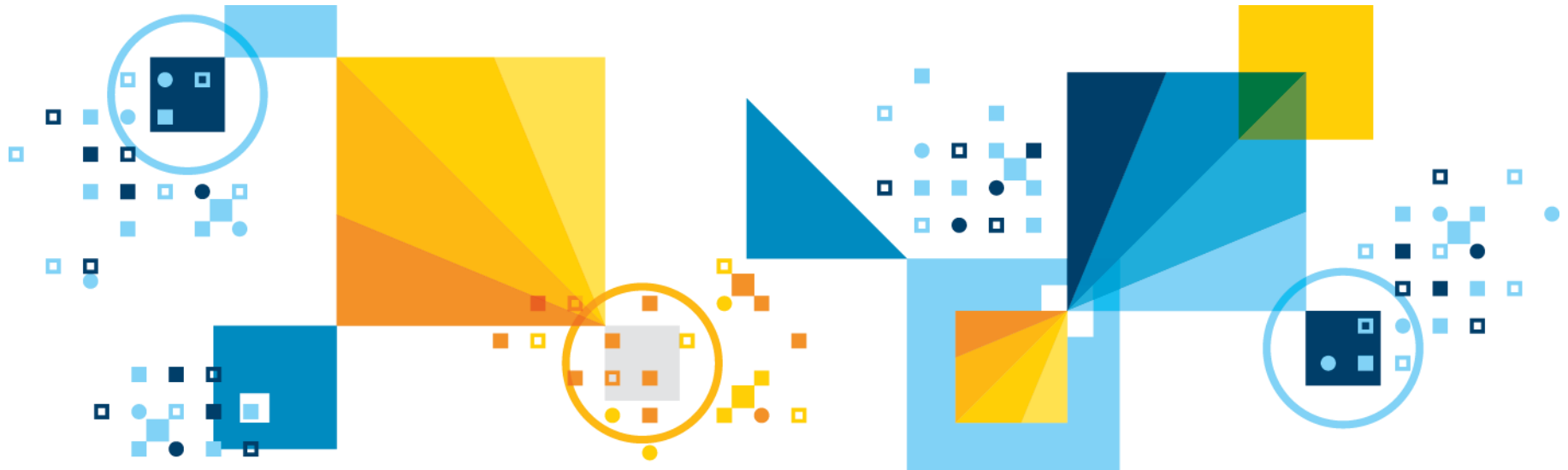


An Overview of CPLEX Mixed Integer Linear Programming Branch-and-Cut

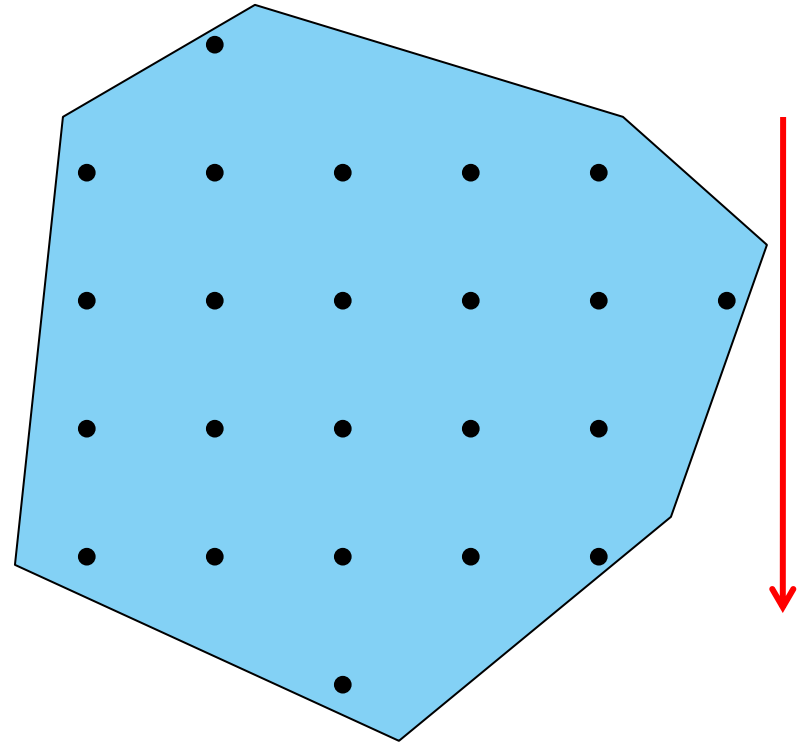
@CPLEX school 2017, Montreal



Mixed Integer linear Programming

- A Mixed Integer (linear) Program (MIP) is a problem of the form

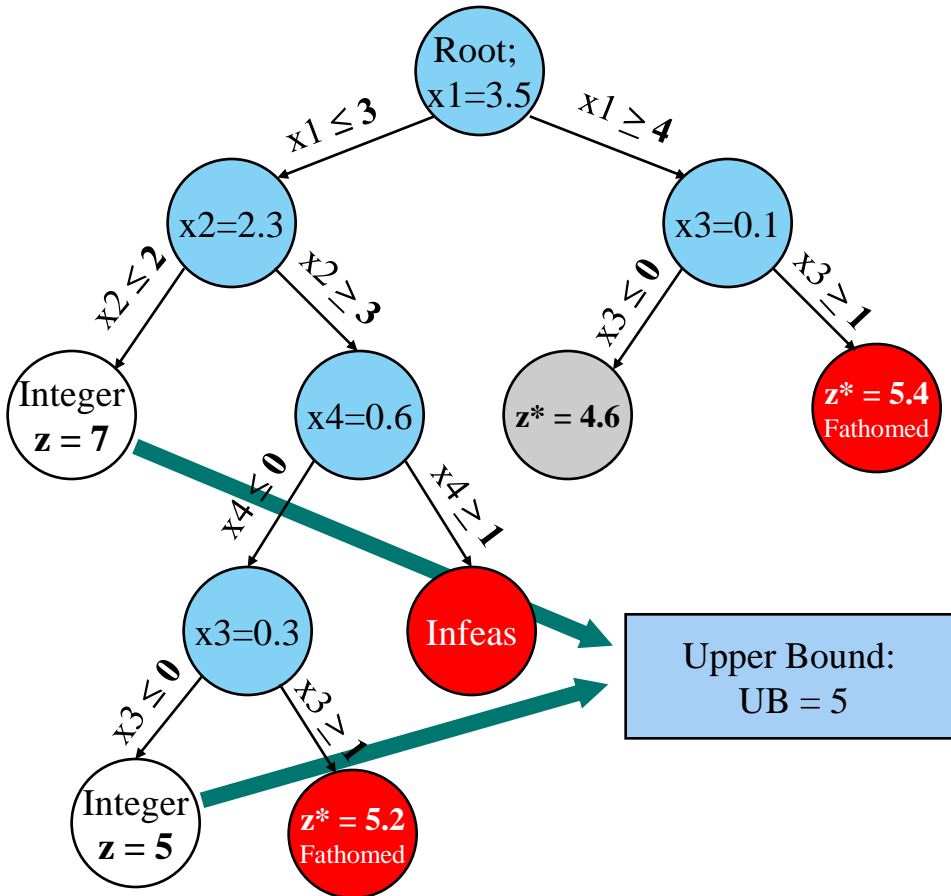
$$\begin{aligned} (MIP) \quad & \text{Minimize} \quad z = c^T x \\ & \text{Subject to} \quad Ax = b \\ & \quad \quad \quad l \leq x \leq u \\ & \quad \quad \quad \text{some or all } x_j \text{ integer} \end{aligned}$$



Some MIP real-world applications

MANUFACTURING	TRANSPORTATION & LOGISTICS	UTILITIES, ENERGY & NATURAL RESOURCES	TELECOM	MULTIPLE/ OTHER
<ul style="list-style-type: none"> • Inventory optimization 	<ul style="list-style-type: none"> • Depot/warehouse location 	<ul style="list-style-type: none"> • Supply portfolio planning 	<ul style="list-style-type: none"> • Network capacity planning 	<ul style="list-style-type: none"> • Workforce scheduling
<ul style="list-style-type: none"> • Supply chain network design 	<ul style="list-style-type: none"> • Fleet assignment 	<ul style="list-style-type: none"> • Power generation scheduling 	<ul style="list-style-type: none"> • Routing 	<ul style="list-style-type: none"> • Advertising scheduling
<ul style="list-style-type: none"> • Production planning 	<ul style="list-style-type: none"> • Network design 	<ul style="list-style-type: none"> • Distribution planning 	<ul style="list-style-type: none"> • Adaptive network configuration 	<ul style="list-style-type: none"> • Marketing campaign optimization
<ul style="list-style-type: none"> • Detailed scheduling 	<ul style="list-style-type: none"> • Vehicle & container loading 	<ul style="list-style-type: none"> • Water reservoir management 	<ul style="list-style-type: none"> • Antenna and concentrator location 	<ul style="list-style-type: none"> • Revenue/Yield management
<ul style="list-style-type: none"> • Shipment planning 	<ul style="list-style-type: none"> • Vehicle routing & delivery scheduling 	<ul style="list-style-type: none"> • Mine operations 	<ul style="list-style-type: none"> • Equipment and service configuration 	<ul style="list-style-type: none"> • Appointment & field service scheduling
<ul style="list-style-type: none"> • Truck loading 	<ul style="list-style-type: none"> • Yard, crew, driver & maintenance scheduling 	<ul style="list-style-type: none"> • Timber harvesting 		
<ul style="list-style-type: none"> • Maintenance scheduling 	<ul style="list-style-type: none"> • Inventory optimization 			<ul style="list-style-type: none"> • Combinatorial auctions for procurement

Core of state of the art MIP solvers: LP-based Branch and Bound (B&B)



- B&B algorithm:
 - Enumerative solution scheme based on the LP relaxation of MIP
- **Bounding:**
 - Nodes with $z^* \geq \text{UB}$ can be **fathomed** without further ramification.
- Key points to avoid exponential explosion of B&B tree:
 - Strong LP relaxation
 - Effective branching rules
 - Primal heuristics

Agenda

- Main building blocks of state of the art MIP solvers
 - Presolve and probing
 - Cutting planes
 - Branching
 - Primal heuristics

- Performance analysis
 - Performance impact of main building blocks

Presolve

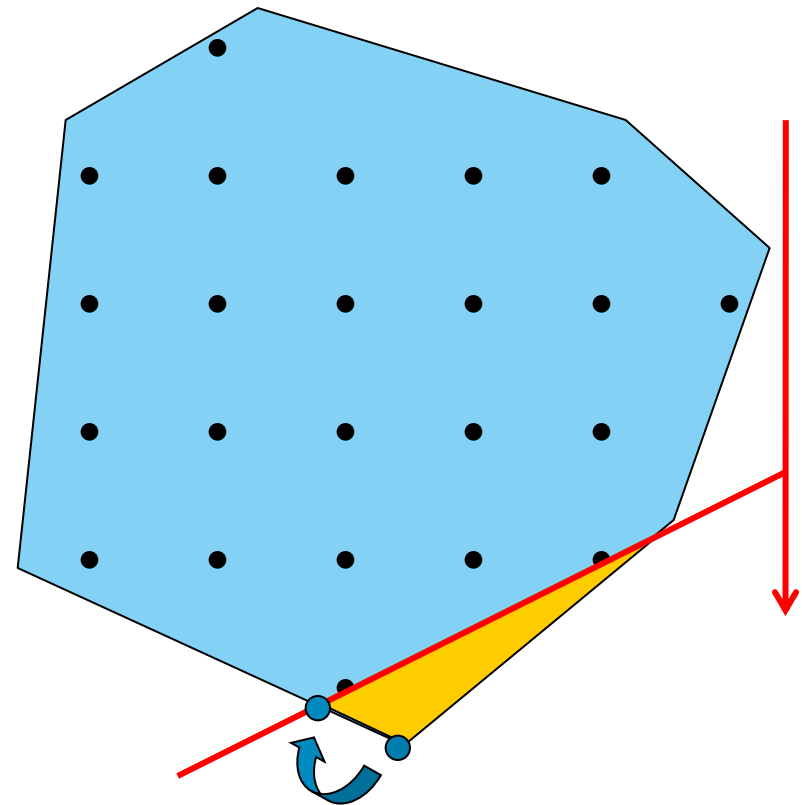
- **M.W.P. Savelsbergh**, Preprocessing and probing techniques for mixed integer programming problems, ORSA Journal of Computing 6, 445-454 (1994)
- Transform a problem **P** to a different but equivalent problem **P'**.
 - **Reduce problem size**
 - Speed-up linear algebra during the solution process
 - **Strengthen the LP relaxation**
 - Identify problem **sub-structures**
 - Cliques, implications, networks, disconnected components, ...
- Primal reductions
 - Preserve the set of feasible solutions
 - Bound strengthening, coefficient strengthening, lifting of constraints, aggregation of variables, detection of implied integer variables and implied continuous variables, ...
- Dual reductions
 - Preserve optimality, but can eliminate feasible and even optimal solutions
 - Dual fixings, fixing and aggregations based on symmetry, removal of parallel or dominated columns, detection of implied integer variable, ...

Probing

- Tentatively set binary variable x to 0 and 1 and propagate fixing
- Inspect implications of $x = 0$ and $x = 1$ to derive globally valid information
 - detection of infeasibility and global fixing of probing variable:
 - $x = 0$ infeasible $\Rightarrow x = 1$
 - global fixings and bound strengthening for implied variables:
 - $x = 0 \rightarrow y \leq u^0, x = 1 \rightarrow y \leq u^1 \Rightarrow y \leq \max\{u^0, u^1\}$
 - aggregations:
 - $x = 0 \rightarrow y = l_y, x = 1 \rightarrow y = u_y \Rightarrow y = l_y + (u_y - l_y) x$
 - implications and cliques:
 - $x = 0 \rightarrow y \leq u^0 \Rightarrow$ store implication in implication (y non-binary) or clique (y binary) table
 - lifting:
 - $ay \leq b, x = 1 \rightarrow ay \leq b - d \Rightarrow ay + dx \leq b$ is valid and dominates $ay \leq b$ (for $d > 0$)
- Applied during and after presolve, during root cut loop, and in node presolve
- Can be very time consuming but also very powerful
 - need to have good dynamically adjusted work limits

Cutting planes

- Valid inequalities for the MIP that cut off integer infeasible points of the LP relaxation
 - Iteratively separated on the fly to **strengthen the LP relaxation**
 - Separated:
 - At the root node (more aggressively)
 - In the tree (less aggressively)
 - Separation must be combined with **clever cut filtering and cut purging** to avoid
 - Numerical difficulties
 - Node throughput slowdown



The root cut loop in CPLEX

```
x* := optimal solution of the LP relaxation
while (x* not integer and cuts seem effective) {
  heuristics, probing, other secret stuff;
  cut separation;
  cut selection/filtering;
  reoptimization;
  cut purging;
}
```

- Cut separation

Several families of cuts are separated at the same time.

- Cut filtering (inspired to Andreello et al., 2007)

Only some of the separated cuts are selected and added to the current formulation:

- Efficacy
- Orthogonality

- Cut purging

After reoptimization, some previously selected cuts may be deemed ineffective and may be discarded.

Cutting planes overview

- General purpose cutting planes in CPLEX
 - Gomory Mixed Integer (GMI) cuts
 - Mixed Integer Rounding (MIR) cuts
 - Lift and Project (L&P) cuts
 - Zero-half cuts
 - Flow cover cuts

- Structural cutting planes in CPLEX
 - Knapsack cover cuts
 - GUB cover cuts
 - Clique cuts
 - Implied bound cuts
 - Multi Commodity Flow (MCF) cuts
 - Flow path cuts

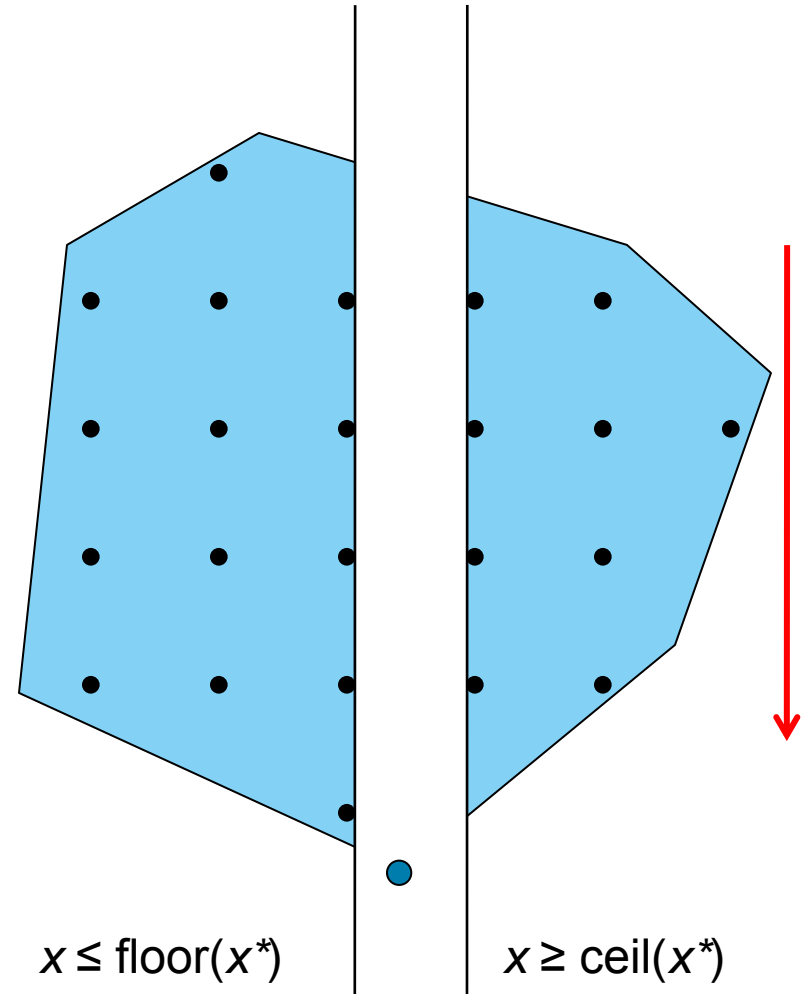
- Survey papers
 - H. Marchand, A. Martin, R. Weismantel, L. Wolsey, Cutting planes in integer and mixed integer programming, *Discrete Applied Mathematics* 123, 397-446, 2002
 - G. Cornuéjols, Valid inequalities for mixed integer linear programs, *Mathematical Programming* 112, 3-44, 2008

Separation of general purpose cutting planes

- Generate a **base inequality** by aggregating constraints of LP relaxation
- Apply **rounding formulas** to aggregated inequality to get a cut
- GMI cuts (Gomory, 1960)
 - Base inequality readily available from the **current tableau**
 - Resulting cut is violated by construction
- MIR cuts (Nemhauser & Wolsey, 1988)
 - Base inequalities **heuristically generated** (CPLEX implementation inspired to Marchand & Wolsey, 2001)
 - Resulting cut may be not violated
- L&P cuts (Balas, 1979, Balas et al., 1993)
 - **Solve Cut Generating LP (CGLP)** to get a tableau different from the one readily available
 - **Separate a GMI** cut from the different tableau
 - Resulting cut is violated by construction is CGLP succeeds finding the alternative tableau
 - CPLEX implementation inspired to Bonami (2012)
- They are “just” alternative strategies for separating **split cuts** (Cook et al., 1990)
 - GMI closure = MIR closure = Split closure (see e.g., Nemhauser & Wolsey, 1990)

Branching

- Divides the feasible region in a manner that all integer feasible solutions belong to one of the branches
 - Standard B&B:
 - up and down **branch on integer variables**
 - Branching on general disjunctions, e.g.:
 - Owen & Mehrotra (2001)
 - Mahajan & Ralphs (2009)
 - Karamanov & Cornéjols (2011)



A generic branching rule

- For each fractional variable $x = x^*$, compute
 - Down score $D(x)$
 - impact of branching down on $x \leq \text{floor}(x^*)$
 - Up score $U(x)$
 - impact of branching up on $x \geq \text{ceil}(x^*)$
 - Overall score $S(x) = f(D(x), U(x))$
 - Variables with large score $S(x)$ are good branching candidates

- **What is an effective rule to compute down and up scores?**
 - Dual bound improvement
 - Child node infeasible or cut-off
 - Help propagation and bound tightening/fixing

- **How to combine down and up score in a single magic number?**
 - $S = \min \{D, U\} + \mu \max \{D, U\}$
 - $S = \max \{D, \epsilon\} * \max \{U, \epsilon\}$
 - More elaborated strategies very recently investigated:
 - Le Bodic & Nemhauser (2015)

Famous branching rules

- Strong branching (Applegate et al., 1995)
 - Limited LP solve for each candidate variable
 - For each fractional variable $x = x^*$, tentatively branch down and up
 - $D(x)$ and $U(x)$ are the improvement in the objective function
 - Can lead to huge reduction in number of nodes
 - But generally too expensive in practice (two limited LP solve for each candidate)

- Pseudo cost branching (Bénichou et al., 1971)
 - Use historical data to predict impact of a branch
 - Record $\Delta obj / \Delta x$ for each branch
 - Maintain $D(x)$ and $U(x)$ as average of recorded values

Famous branching rules

- Pseudo cost with strong branching initialization (Linderoth & Savelsbergh, 1999)
 - If pseudo cost not available, initialize it with strong branching.

- Reliability branching (Achterberg et al., 2005)
 - Consider pseudo costs on **x** reliable only if **x** has been branched on **r** times
 - Among fractional variables, identify the ones with unreliable pseudo cost
 - Apply strong branching to some (possibly, all) unreliable candidates to update their pseudo cost and make them reliable
 - Apply **pseudo cost branching** to all candidates with **reliable pseudo cost**

Branching in CPLEX

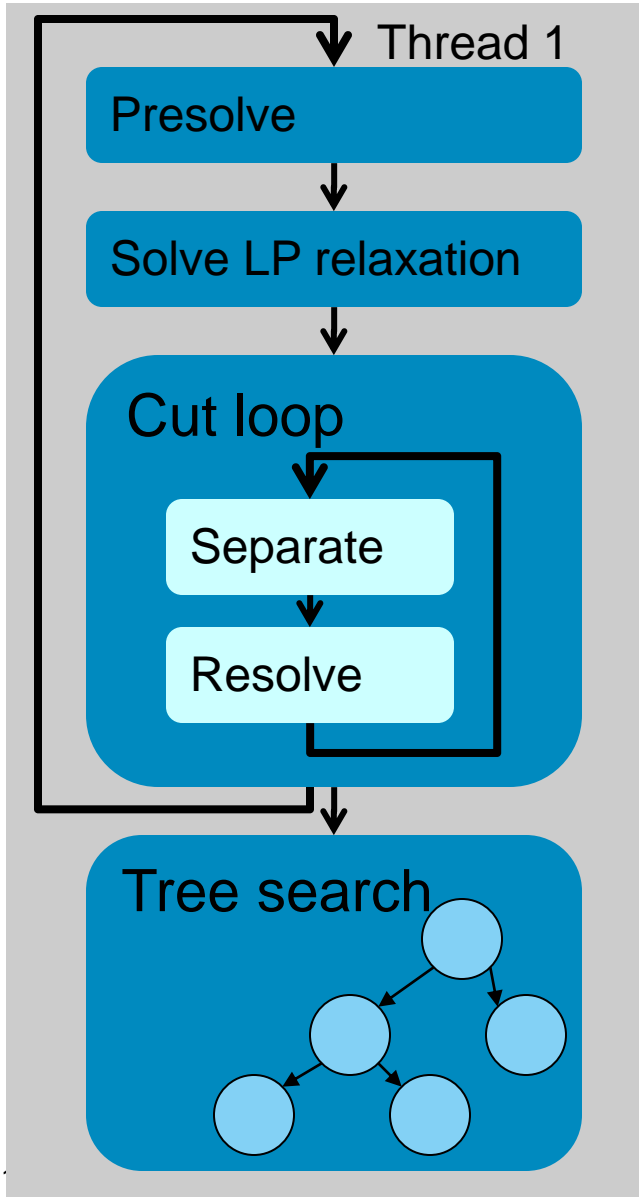
- Default CPLEX strategy is similar to Hybrid branching (Achterberg and Berthold, 2009):
 - Reliability branching
 - Conflict scores
 - Down and up score based on conflict table
 - Idea: prefer branching candidates that are more likely to yield infeasible nodes after branching.
 - Pseudo reduced cost branching (similar to Patel and Chinneck, 2007)
 - Estimate branching impact on the dual bound from the dual solution
 - Inference scores
 - Down and up scores based on clique table and implication table
 - Idea: prefer branching candidates that allow more propagation

Primal heuristics

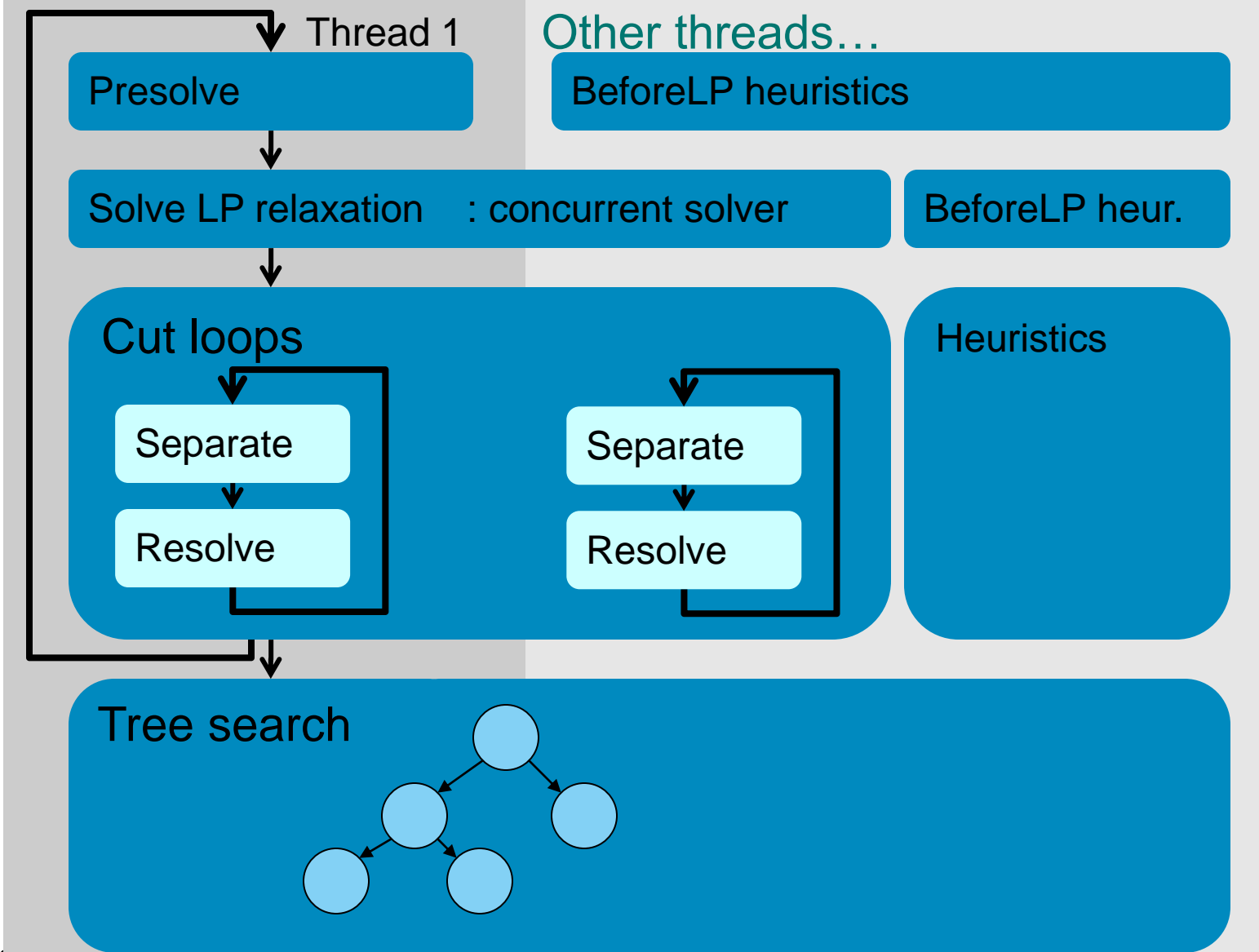
- Starting heuristics
 - Heuristics that do not need any LP solution available
 - before LP heuristics, ...
 - Heuristics based on the current LP solution
 - Diving heuristics
 - Simulate depth-first-search with special branching strategy
 - Propagate and resolve LP
 - ...
- Improving heuristics
 - Heuristics that do not need any LP solution available
 - Neighborhood depends on incumbent solution only
 - Heuristic based on the current LP solution
 - E.g., RINS (Danna et al., 2005)
- Survey paper:
 - M. Fischetti, A. Lodi, Heuristics in mixed integer programming, in J.J. Cochran (ed.) Wiley Encyclopedia of Operations Research and Management Science, Vol. 8, pp. 738-747, John Wiley & Sons, 2011

In Summary

CPLEX MIP Solver

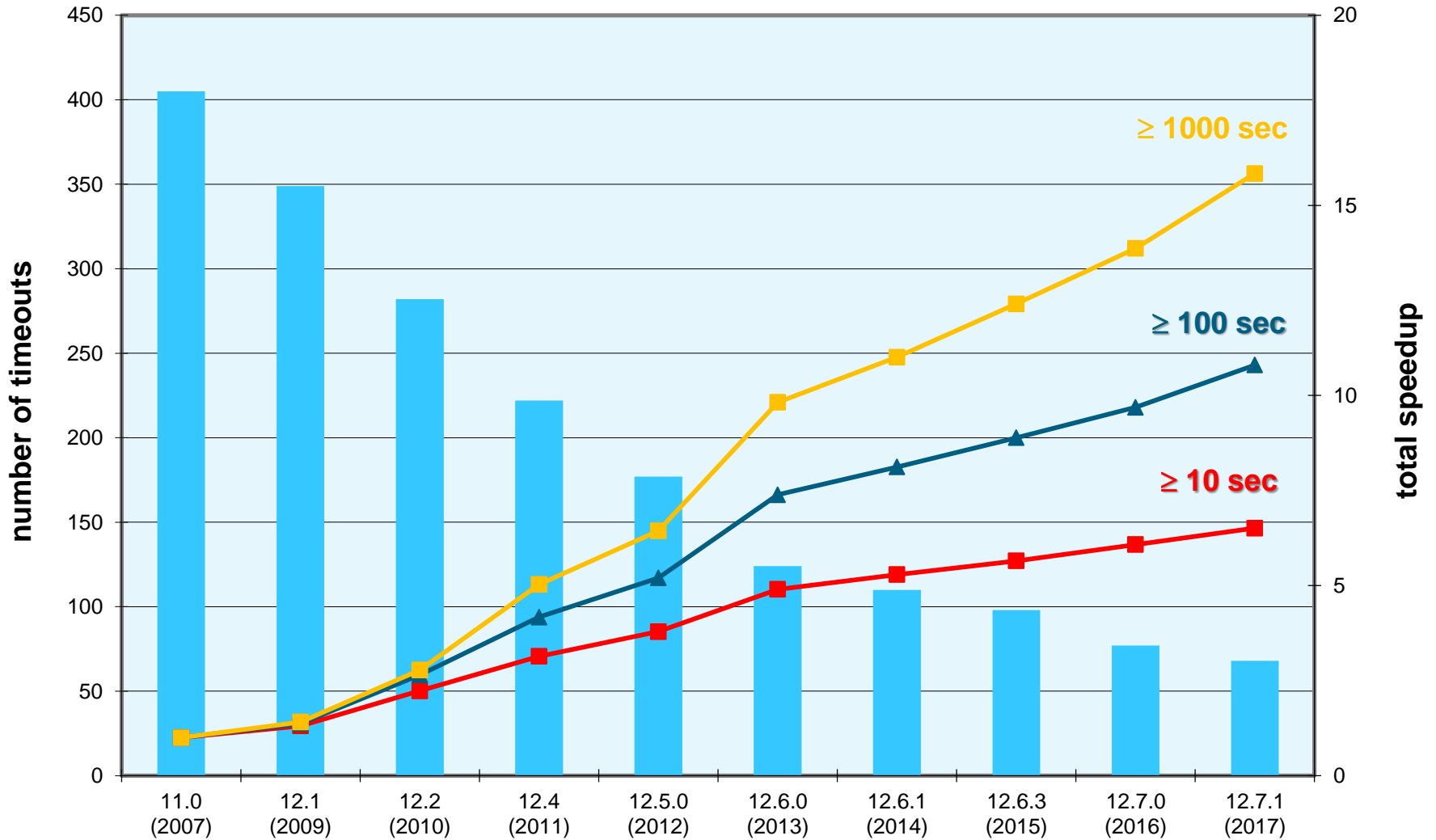


CPLEX parallel MIP Solver



MIP Performance Analysis

CPLEX MILP performance evolution



Date: 18 June 2017
 Testset: MILP: 4009 models
 Machine: Intel X5650 @ 2.67GHz, 24 GB RAM, 12 threads, deterministic
 Timelimit: 10,000 sec

Main building blocks: Measuring performance impact

- How important is each component?

Compare runs with feature turned on and off

- Solution time degradation (geometric mean)
- # of solved models
 - Essential or just speedup?
- Number of affected models
 - General or problem specific?

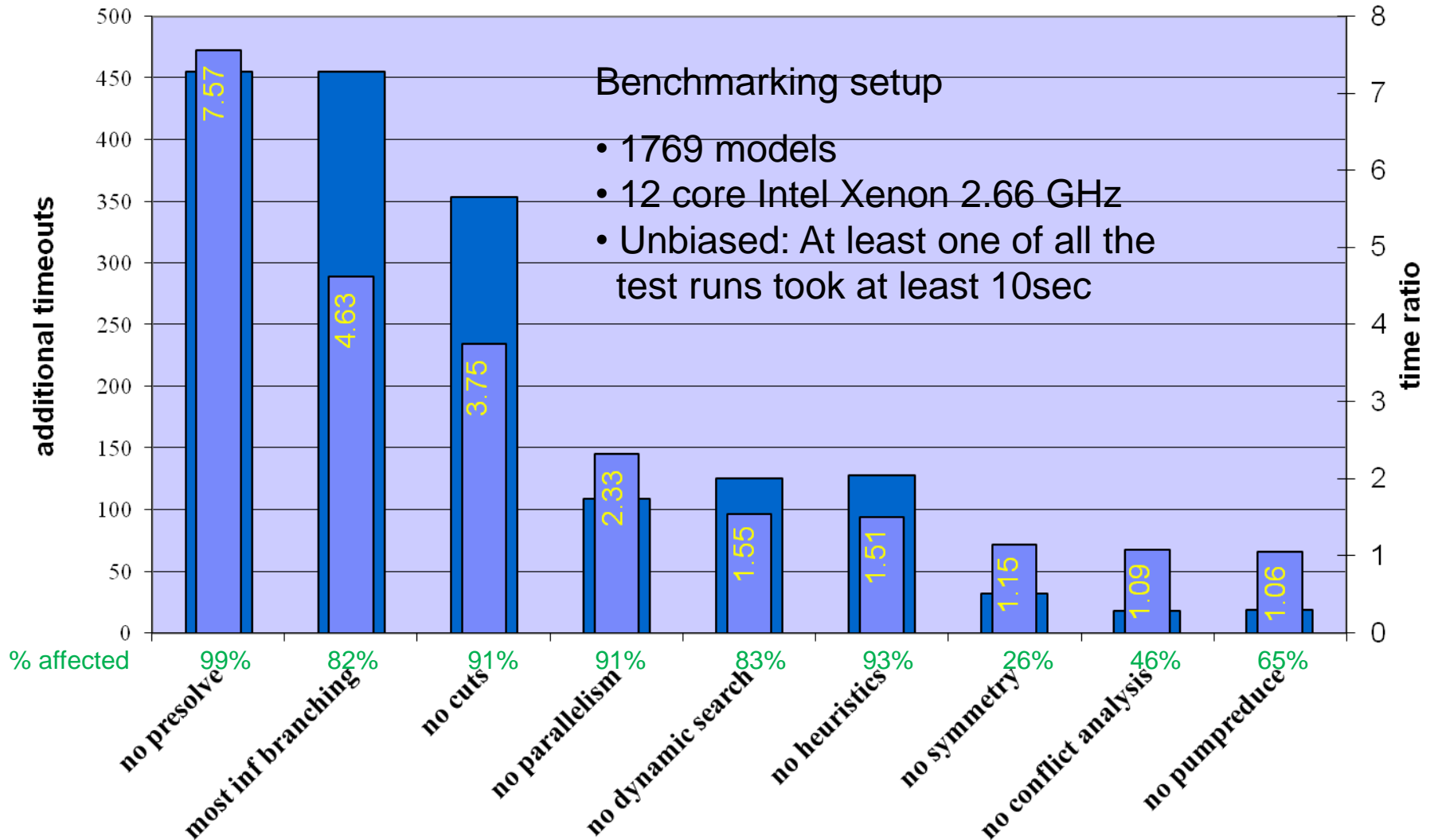
- Experiments conducted with CPLEX 12.5.0 (2012)

- Several features not available yet, e.g.,
 - L&P cuts (added in CPLEX 12.5.1)
 - Parallel cut loop (added in CPLEX 12.5.1)

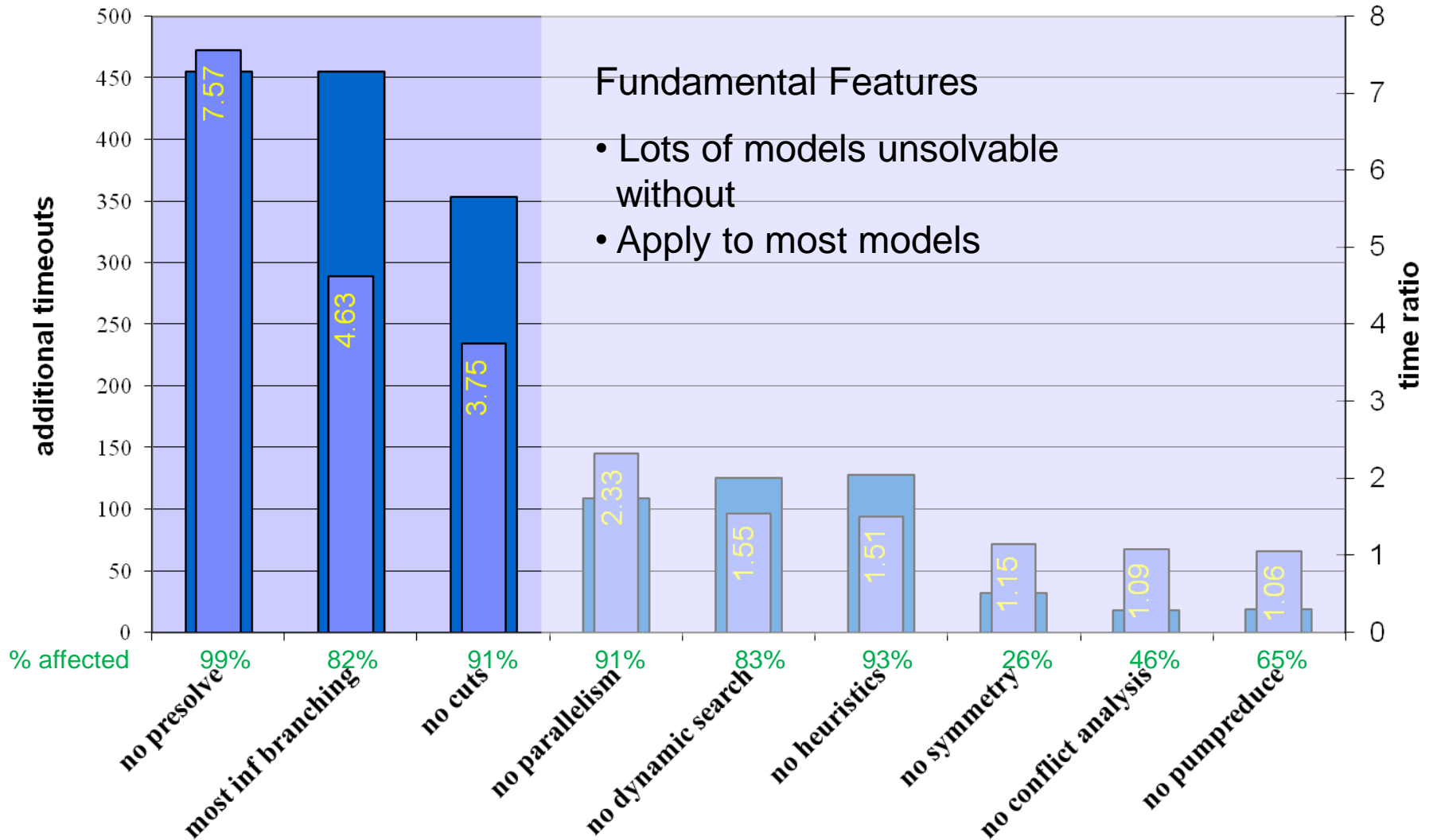
- More detailed analysis in:

T. Achterberg and R. Wunderling, “Mixed Integer Programming: Analyzing 12 Years of Progress”, in: Jünger and Reinelt (eds.) Facets of Combinatorial Optimization, Festschrift for Martin Grötschel, pp.449-481, Springer, Berlin-Heidelberg (2013)

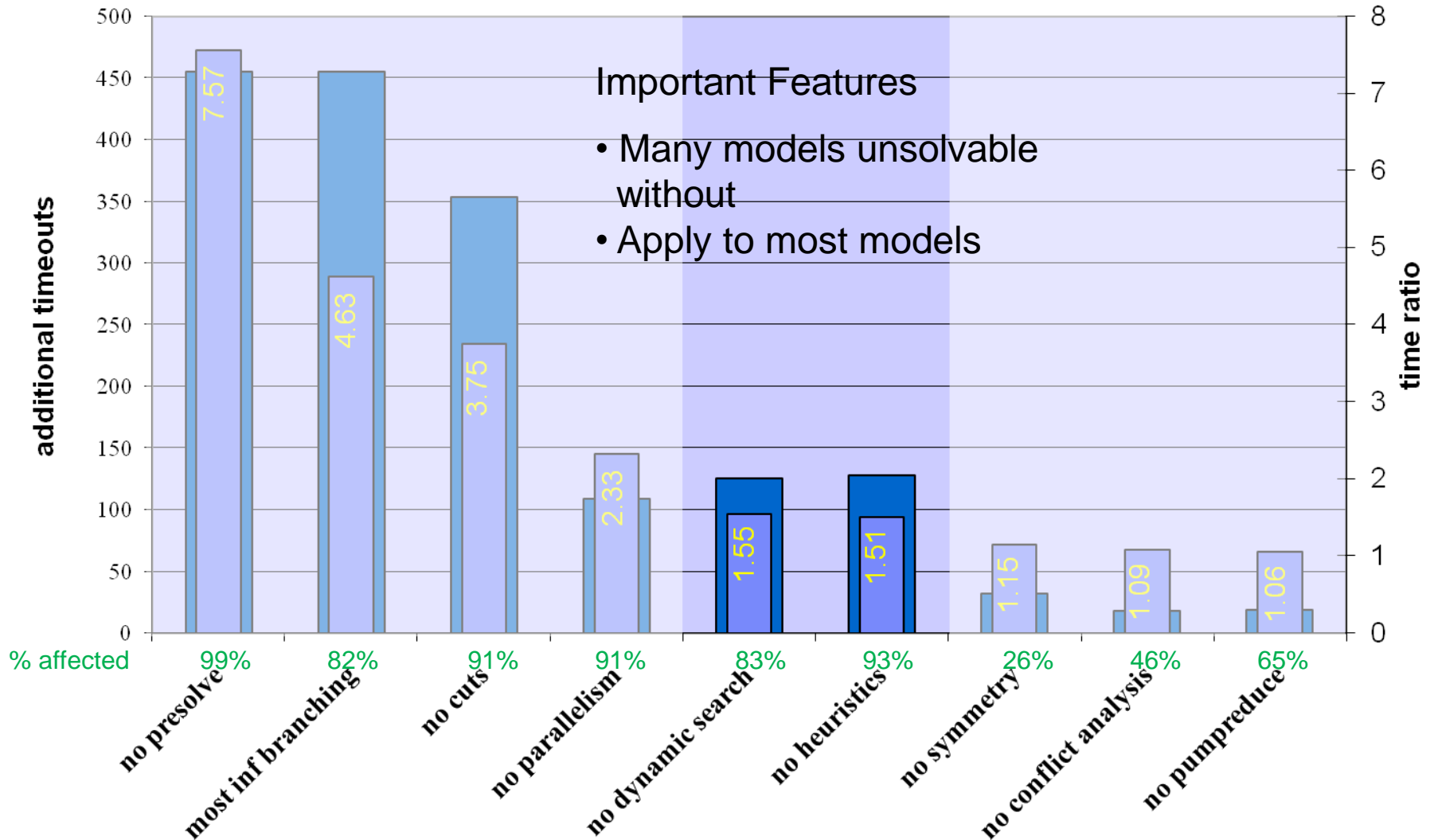
Component Impact CPLEX 12.5.0 - Summary



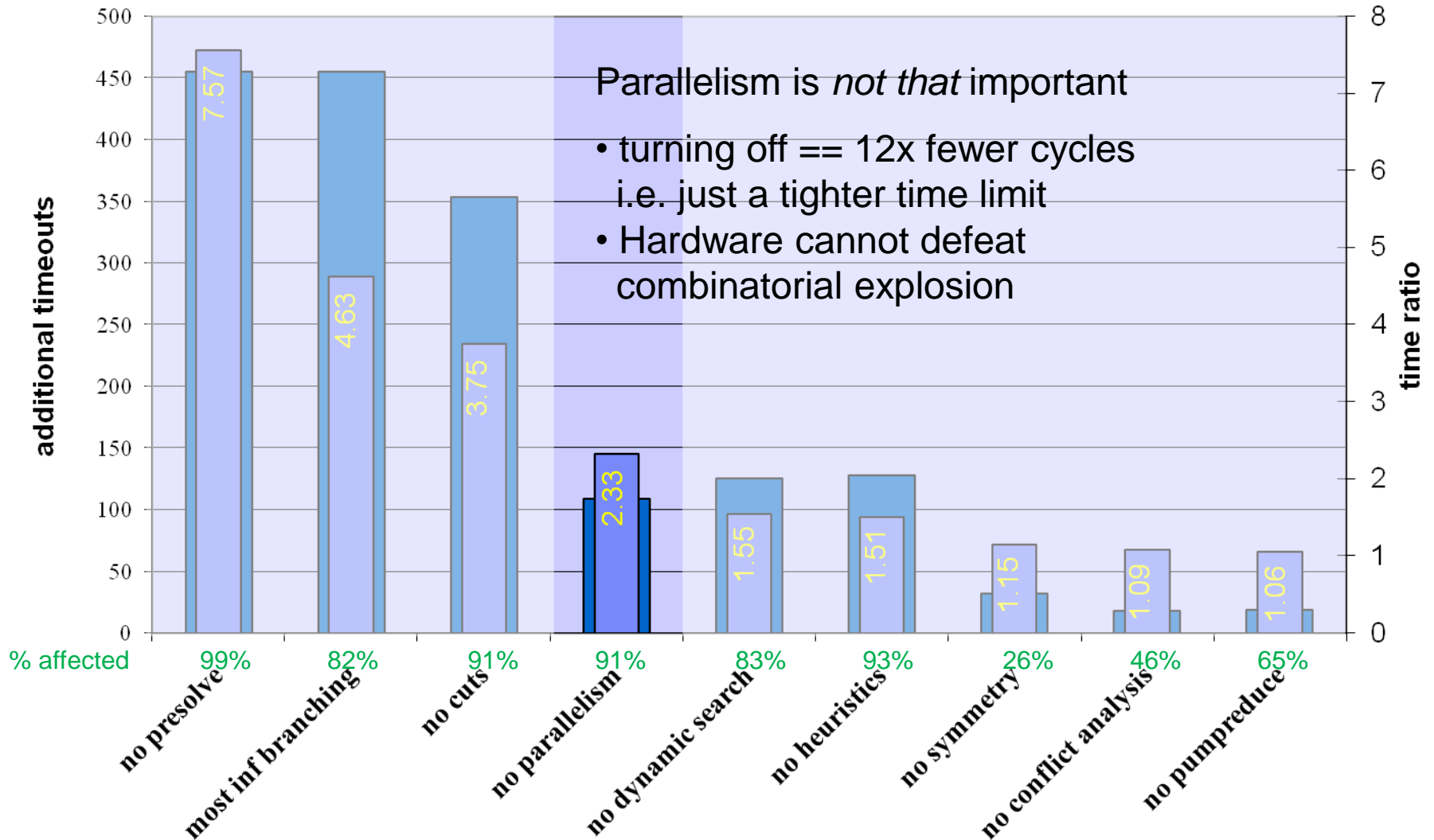
Component Impact CPLEX 12.5.0 - Summary



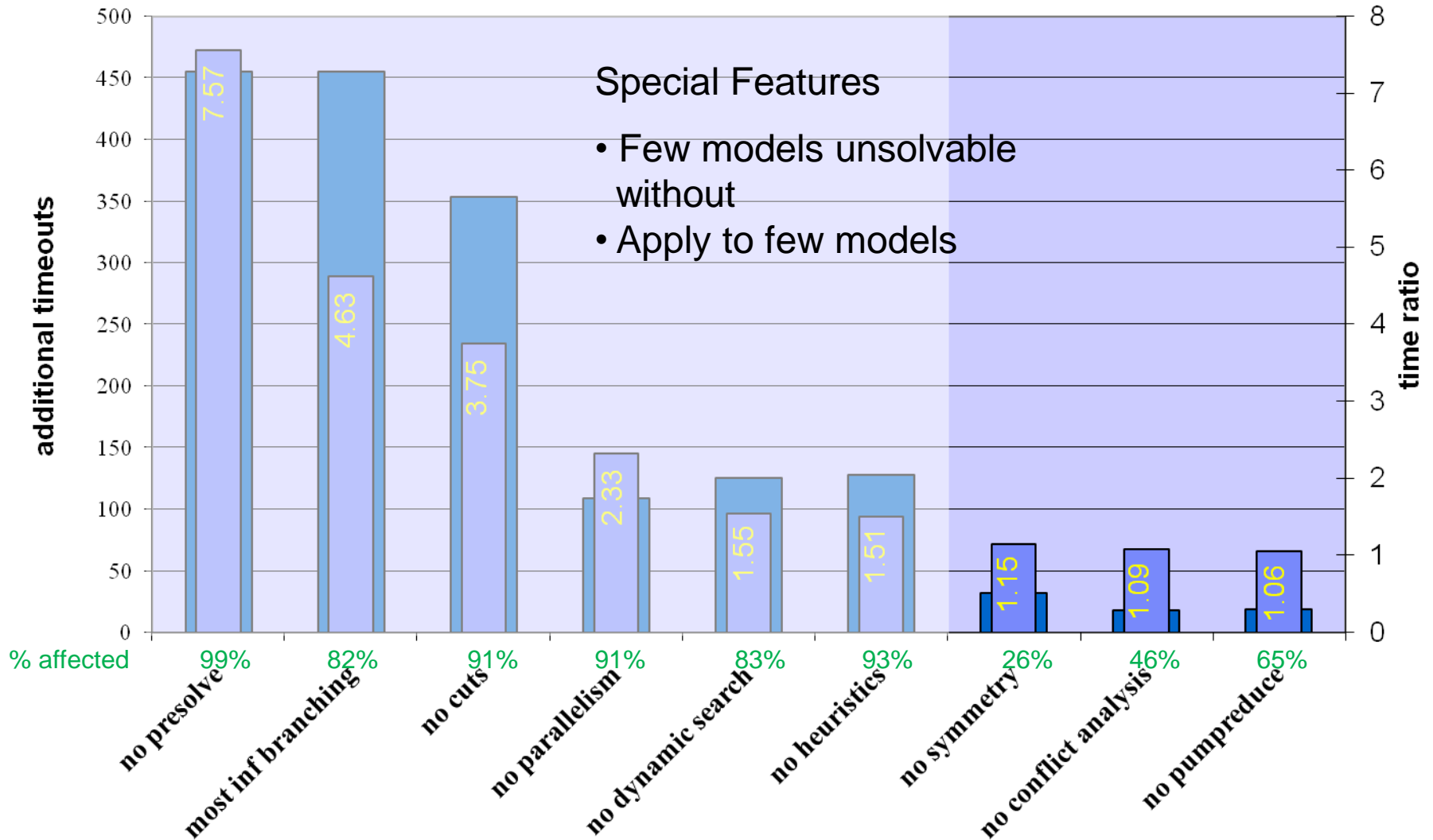
Component Impact CPLEX 12.5.0 - Summary



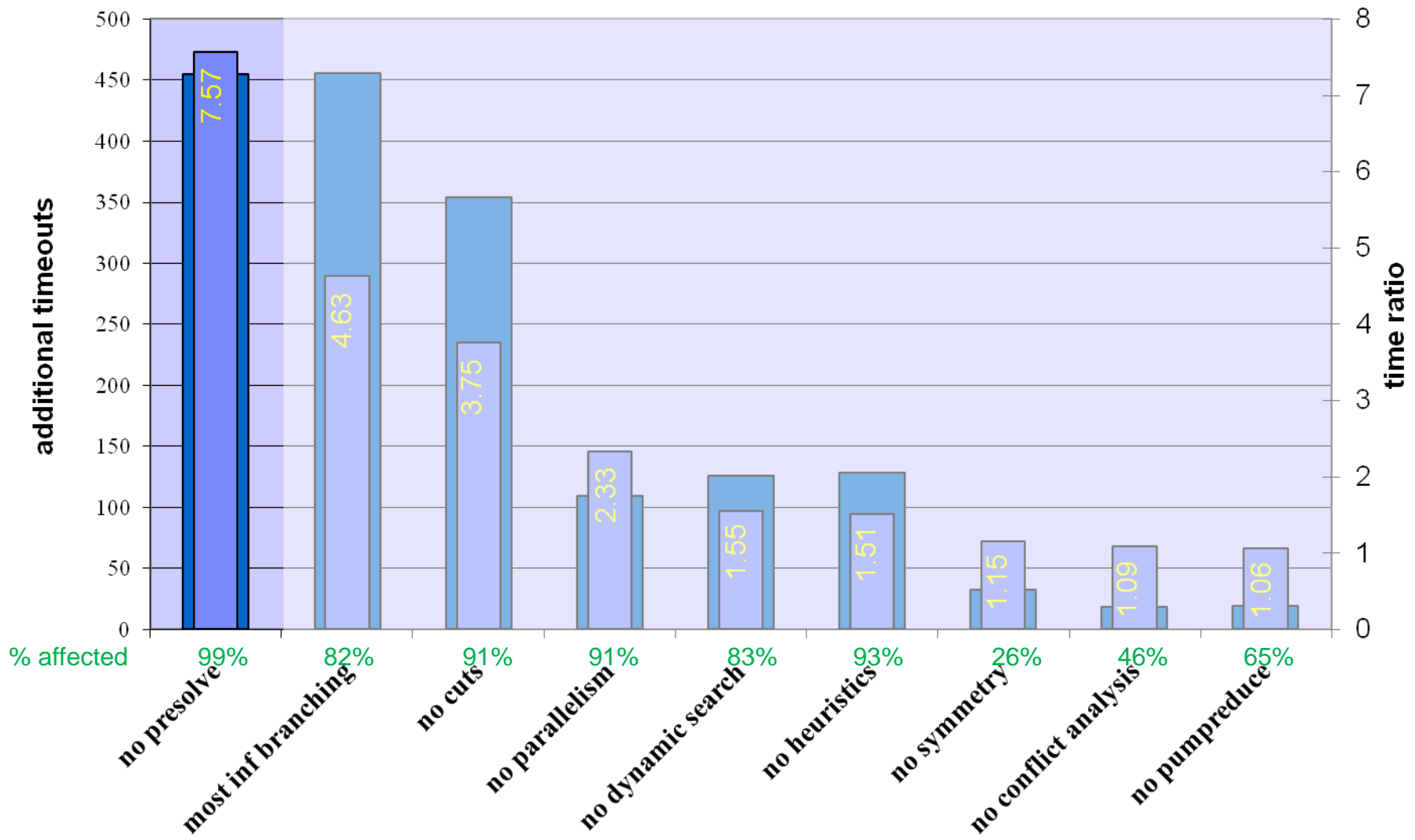
Component Impact CPLEX 12.5.0 - Summary



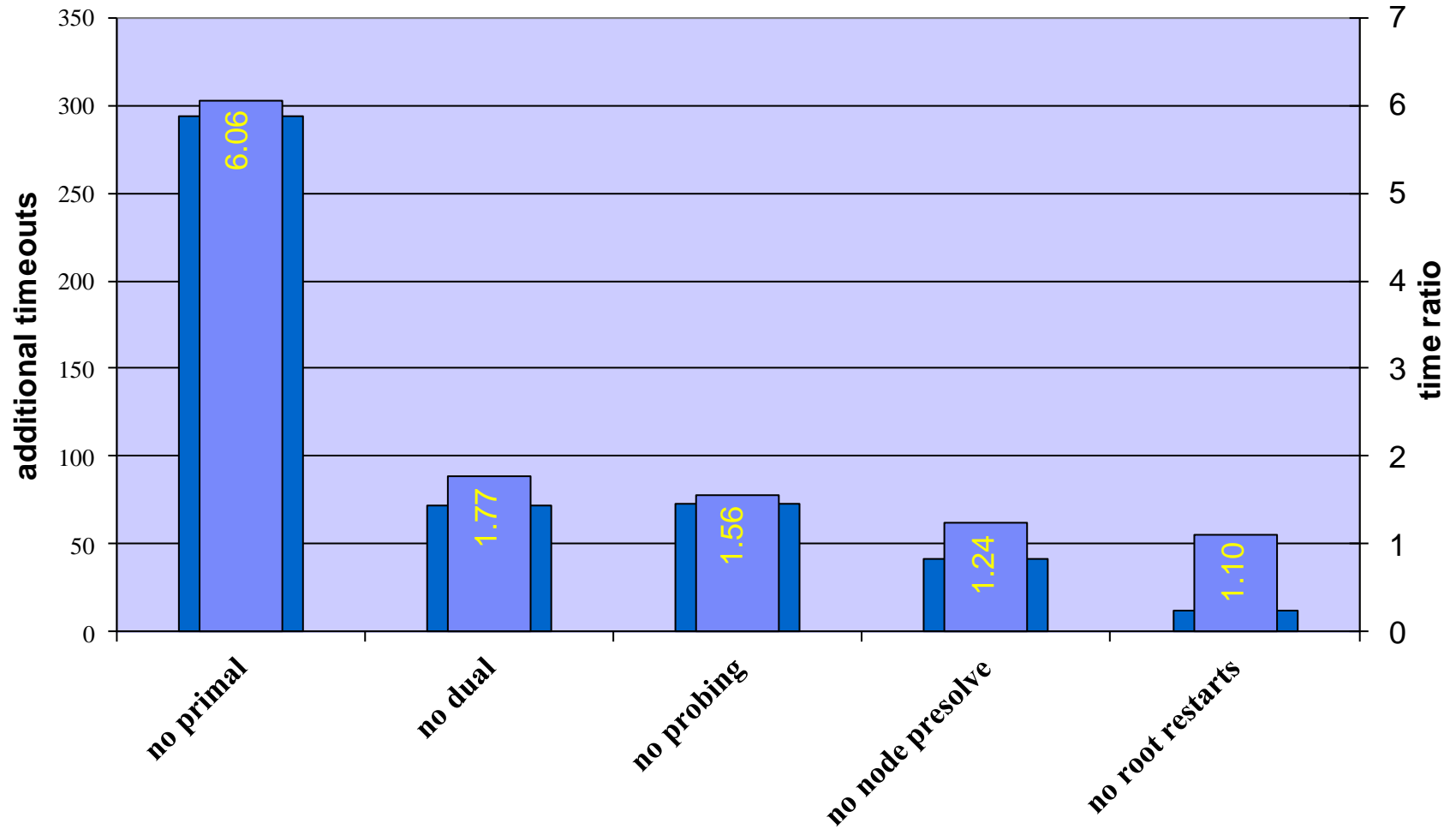
Component Impact CPLEX 12.5.0 - Summary



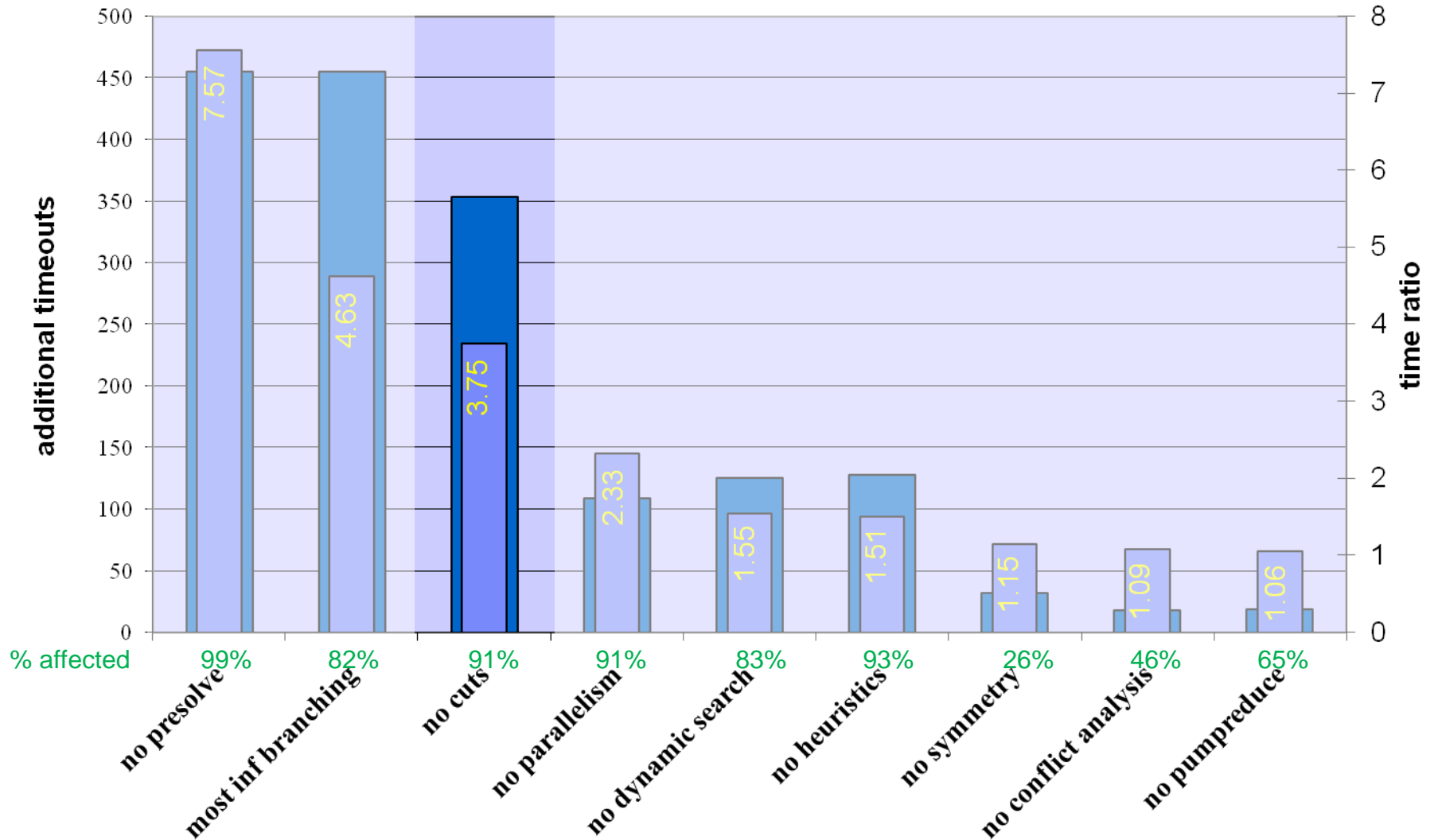
Component Impact CPLEX 12.5.0 - Summary



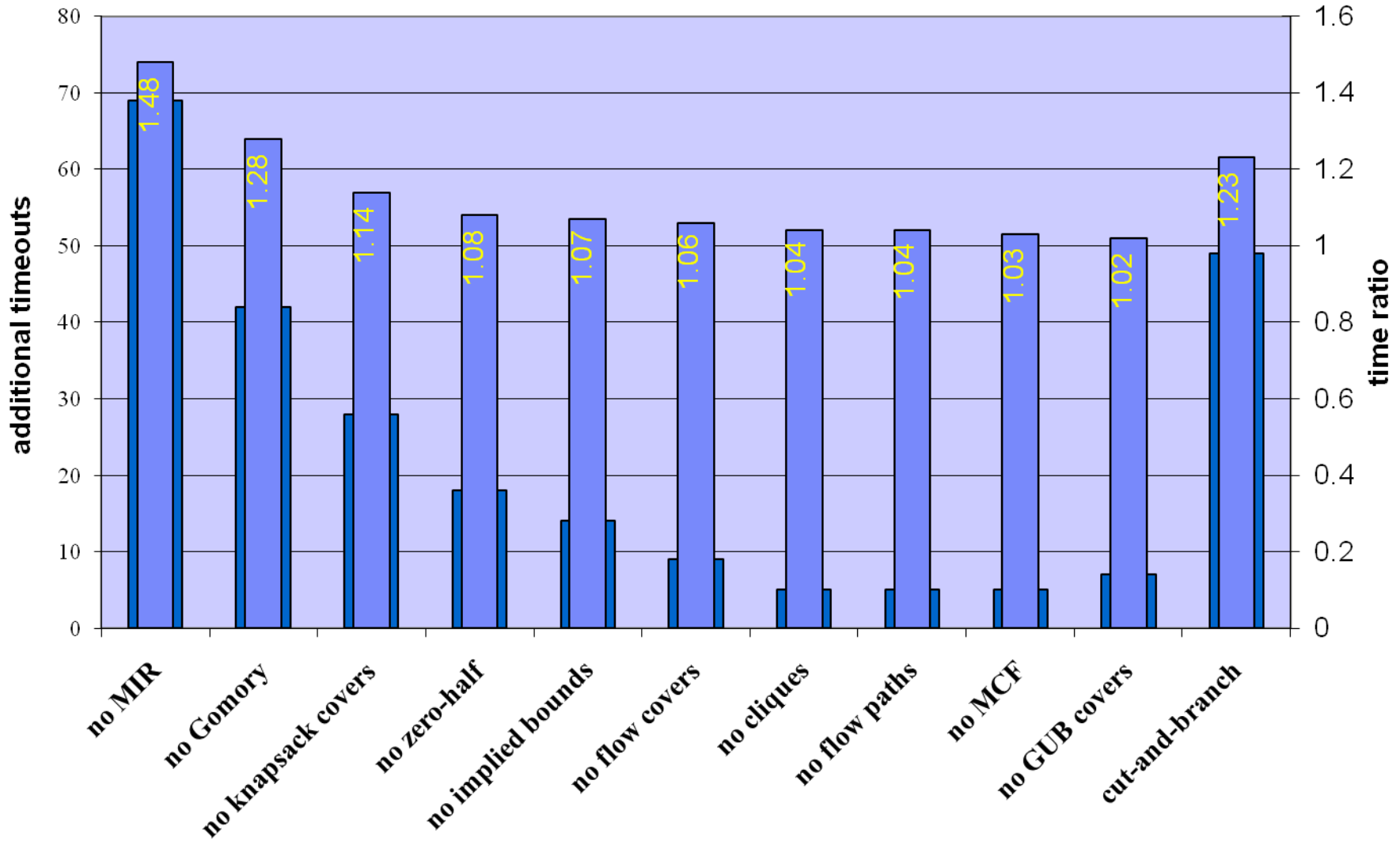
Component Impact CPLEX 12.5.0 – Presolve



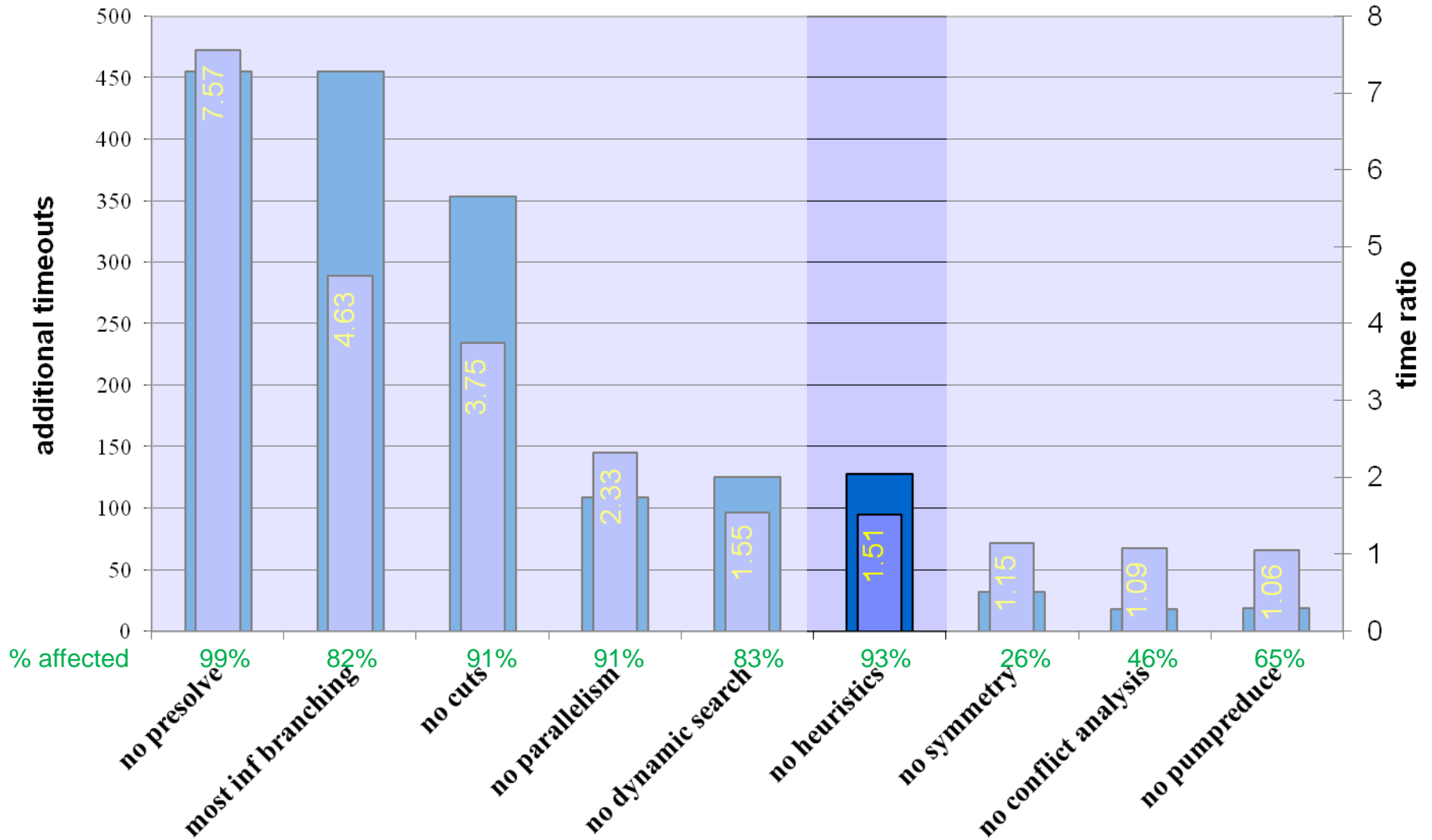
Component Impact CPLEX 12.5.0 - Summary



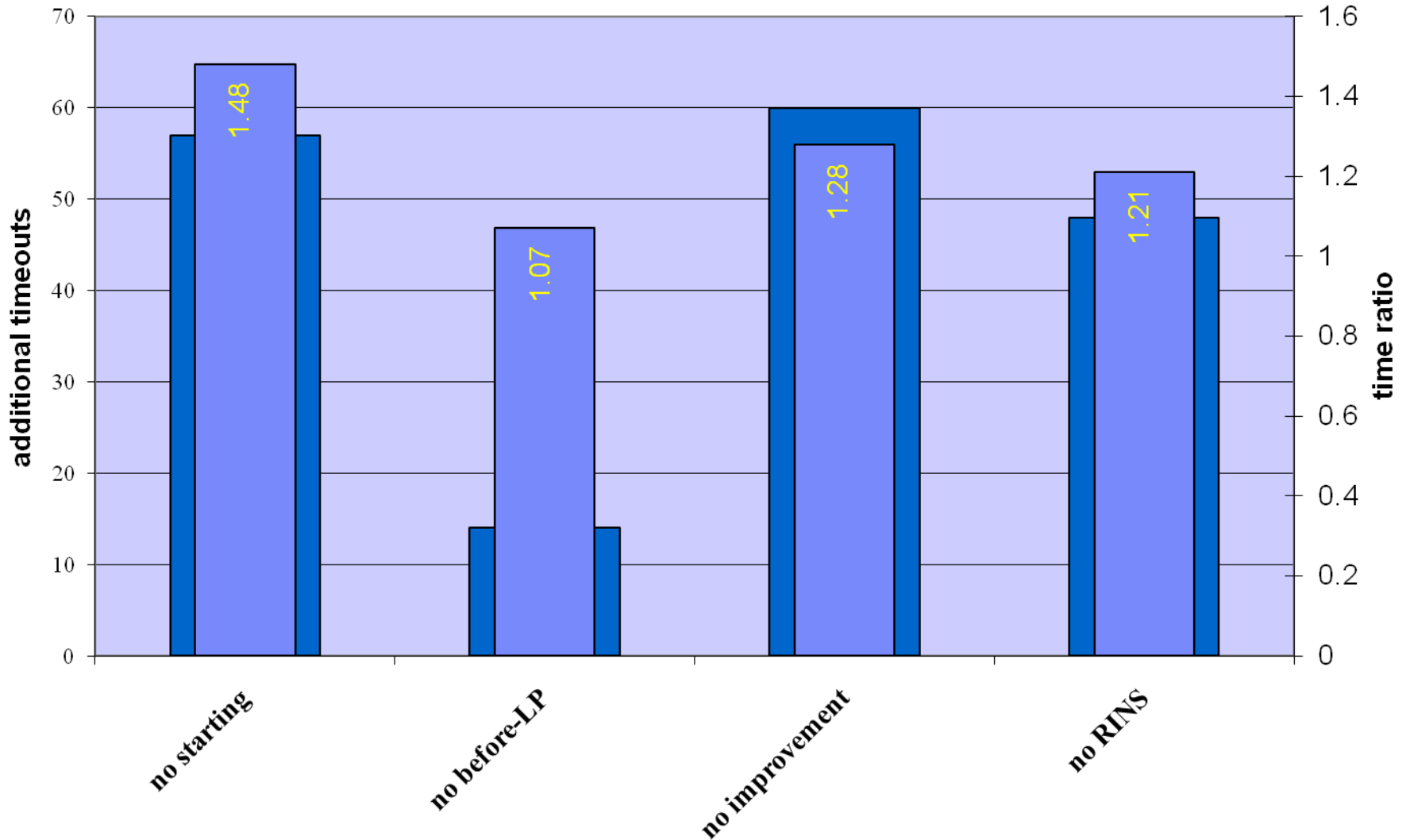
Component Impact CPLEX 12.5.0 – Cutting planes



Component Impact CPLEX 12.5.0 - Summary



Component Impact CPLEX 12.5.0 – Primal heuristics



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