

LOCAL BRANCHING

or

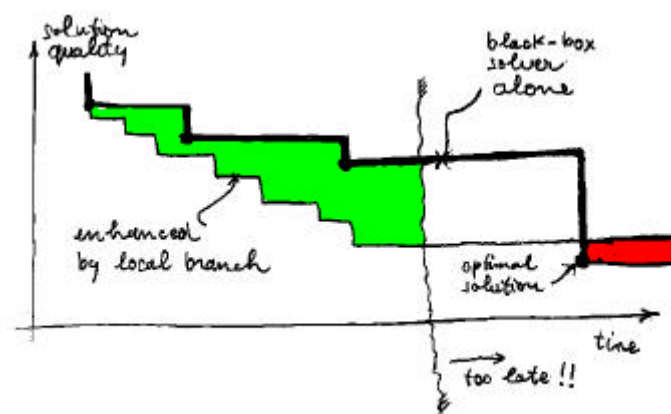
How to enhance the heuristic behaviour of your favourite 0-1 MIP solver

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Presented in AUSSOIS, January 6-11, 2002

0-1 Mixed-Integer Programs

We consider generic Mixed-Integer Linear Programming problems (MIP's) with 0-1 variables

$$\begin{aligned} \min \quad & c^T x \\ & Ax \geq b \\ & x_j \in \{0,1\}, \quad \forall j \in \beta \neq \emptyset \\ & x_j \text{ integer}, \quad \forall j \in \gamma (\supseteq \emptyset) \end{aligned}$$

Relevant cases:

- **0-1 ILP's** (generic or with a special structure)
 - set partitioning/covering models (crew scheduling etc.)
 - TSP, VRP, etc.
- **MIP's with no “general integer” variables**
- **MIP's with both general integer and binary variables**, the latter being often used to activate/deactivate costs/constraints (possibly using BIG-M tricks...)

Assumption: once the binary variables have been fixed, the problem becomes (relatively) easy to solve

Hard-to-solve 0-1 MIP's (in practice)

- In many practical cases, generic 0-1 MIP's can be solved in a satisfactory way by general-purpose commercial software which delivers:
 - Provably optimal solution
 - Heuristic solutions with a practically-acceptable error

Most MIPLib instances are of this type!

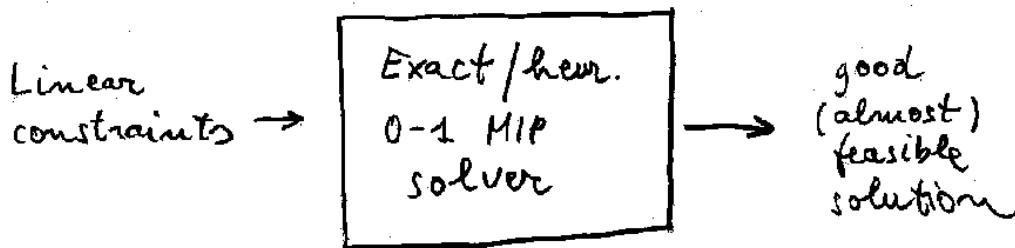
- Unfortunately, in other cases general-purpose software is not adequate and one has to:
 - Play with the MIP solver parameters (“emphasize integrality” etc.) so as to convince the solver to deliver, at least, a good solution
 - Design and use ad-hoc heuristics—thus losing the advantage of working in a generic MIP framework

Many real-world instances are of this type!

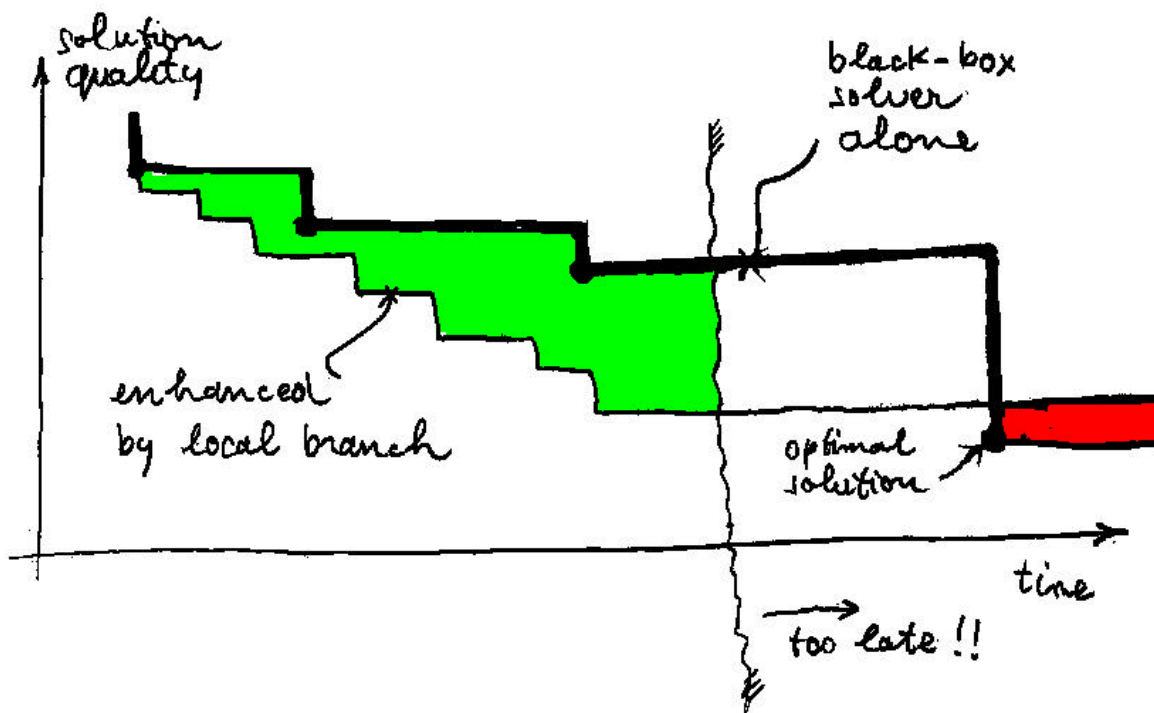
Better heuristics for general 0-1 MIP's strongly required!

A general heuristic framework

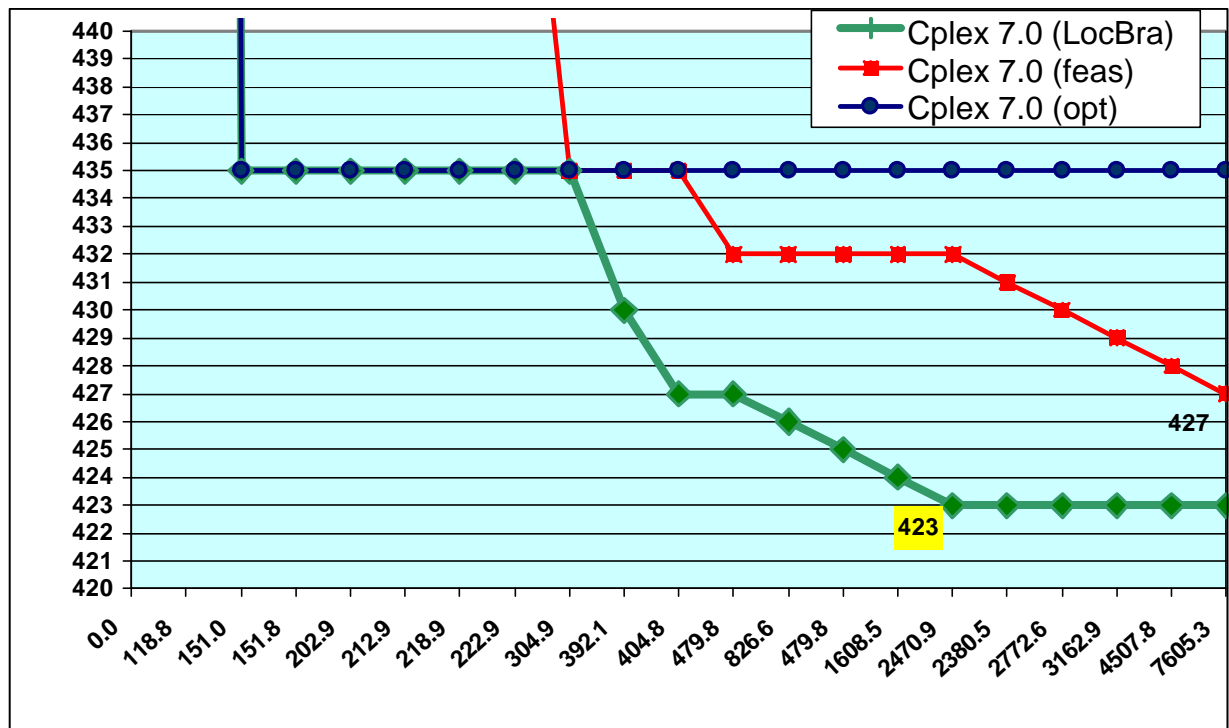
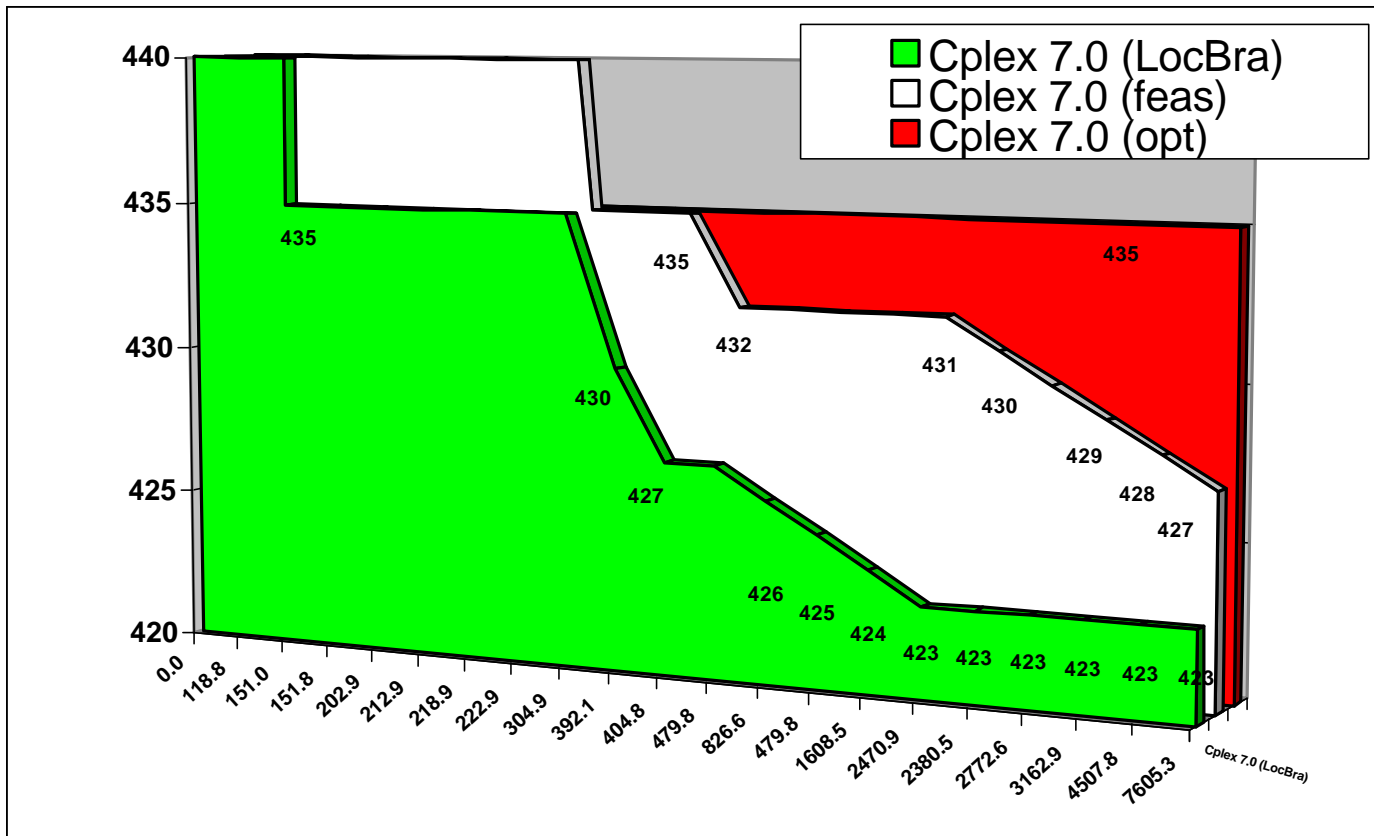
- We aim at embedding a **black-box** (general-purpose or specific) 0-1 MIP solver within an overall **heuristic framework** that “helps” the solver to deliver improved heuristic solutions



The available black-box module



The desired “Italian flag”



The Local Branch heuristic on a hard MIPLIP problem (seymour.lp)

MIPLIP problem seymour.lp

CPLEX 7.0: MIP emphasis: optimality

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- Final Sol. after 18000.0 sec.s : 435.0

CPLEX 7.0: MIP emphasis: integer feasibility

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CPLEX 7.0 & Local Branching (k=10)

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- Local Branch Time = 2470.9 sec. : 423.0

MIPLIP: problem arki001.lp

CPLEX 7.0: MIP emphasis: optimality

➤	Elapsed b&b time =	21.12 sec.	7,594,629.2
➤			7,590,295.2
➤			7,590,247.2
➤	Elapsed b&b time =	212.32 sec.	7,585,194.4
➤	Elapsed b&b time =	1897.90 sec.	7,584,116.1
➤	Elapsed b&b time =	2088.58 sec.	7,583,895.3
➤			7,583,878.4
➤	Elapsed b&b time =	2450.85 sec.	7,582,953.8
➤	Elapsed b&b time =	2613.20 sec.	7,582,840.6
➤	Elapsed b&b time =	4160.22 sec.	7,582,751.4
➤	Elapsed b&b time =	6216.88 sec.	7,582,634.8
➤	Elapsed b&b time =	7161.85 sec.	7,582,414.4
➤	Elapsed b&b time =	7161.85 sec.	7,582,302.6
➤	Elapsed b&b time =	14322.80 sec.	7,582,202.7
➤	Elapsed b&b time =	16237.02 sec.	7,582,031.3
➤	Elapsed b&b time =	16237.02 sec.	7,582,024.4

CPLEX 7.0 & Local Branching (k=10)

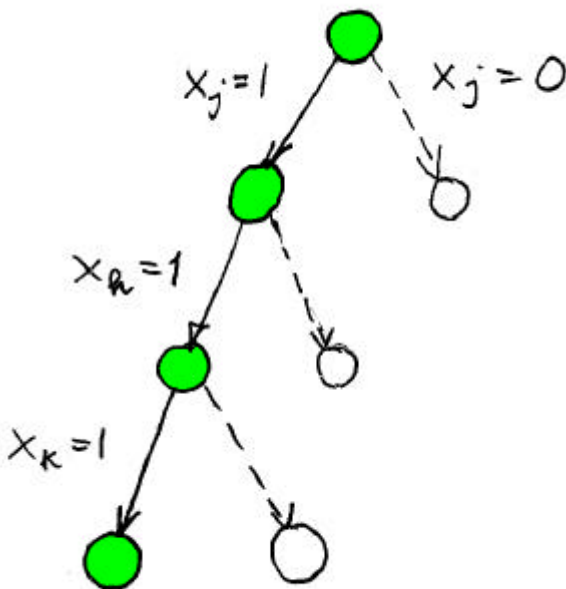
➤	Time =	21.6	best sol =	7,594,629.28
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➤	Time =	2730.1	best sol =	7,580,947.53
➤	Time =	3630.4	best sol =	7,580,937.90
➤	Time =	4080.5	best sol =	7,580,925.20

Variable-fixing strategy (hard version)

- A commonly-used (often quite effective) heuristic framework
- Let x^H be an (almost) feasible “target solution”, and let

$$S = \{j \in B : x_j^H = 1\}$$

denote its **binary support** (binary var.s at value 1).



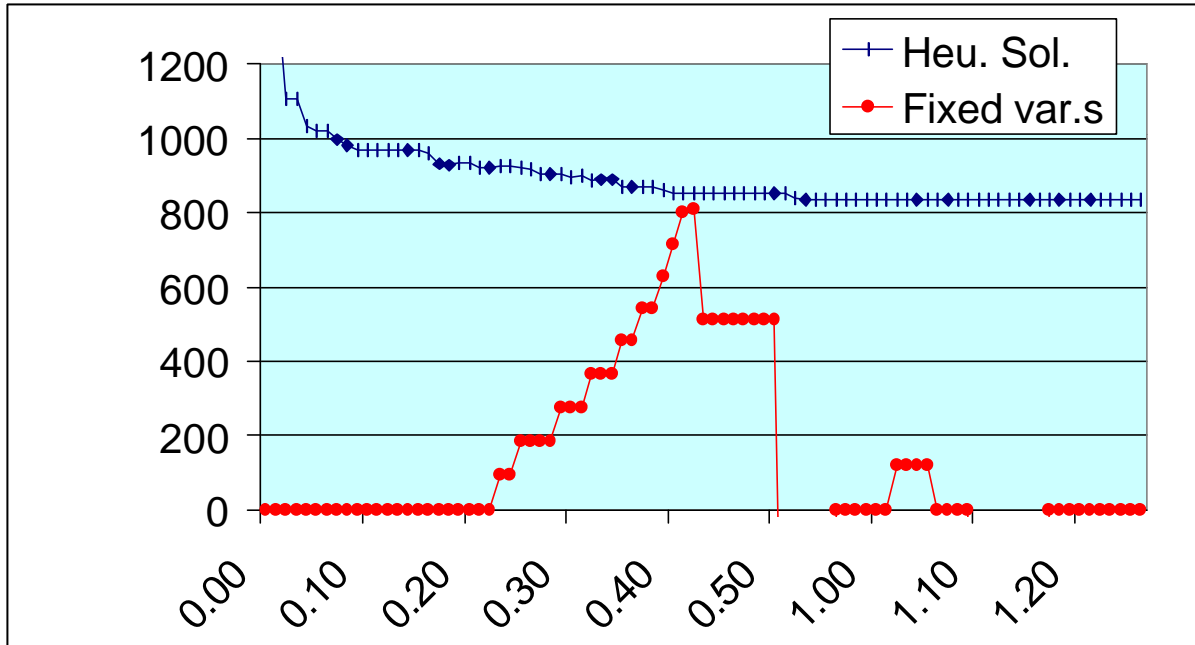
- **Heuristic** depth-first search of the branching tree:
- iteratively **fix to 1** certain “highly efficient” variables in S (**green nodes**)
- apply the **black-box module** to some **green nodes** only
- only limited **backtracking** allowed

Advantages:

- Problem size quickly reduced: the black-box solver can concentrate on smaller and smaller “**core problems**”
- The black-box solver is applied over and over on different subproblems (**diversification**)

Example:

Black-box = Crew scheduling solver based on a set covering model (dynamic column-generation & Lagrangian heuristics)



Crew scheduling heuristic TURNI using variable fixing

Disadvantages:

- How to choose the “highly efficient variables in S ” to be fixed?
- Wrong choices at early levels are typically very difficult to detect, even when **lower bounds** are computed along the way

Feasible solutions only available after several fixings, at a **deep level** in the branching tree

The lower bound does not help at early levels: it often stays quiet for several fixings, and **bumps** suddenly after an apparently innocent late fixing

How to reach a sufficiently-deep branching level with a good lower bound?

Variable-fixing strategy (soft version): local branching

- As before, let x^H be a (almost) feasible “target solution” and denote by $S^H = \{j \in B : x_j^H = 1\}$ its binary support
- Don't decide the actual variables in S^H to be fixed (a difficult task!), but just their **number** $|S^H| - k$
- Introduce the **local branching** constraint

$$x(S^H) := \sum_{j \in S^H} x_j \geq |S^H| - k$$

$$\left(\text{or} \quad \sum_{j \in S^H} (1 - x_j) \leq k \right)$$

so as to define a convenient **k-OPT neighbourhood** $N(x^H, k)$ of the target solution x^H (the larger k , the larger the neighbourhood)

“Akin to k-OPT for TSP”

- Search $N(x^H, k)$ by means of the black-box module
- Repeat with a different target solution (if available) and/or with a modified k (basic idea: to be elaborated in the sequel...)

Conjecture: a small value of k drives the black-box solver towards integrality as effectively as fixing a large number of variables, but with a **much larger degree of freedom** → better solutions can be found.

Local branching in an exact solution framework

- **Black-box module** = generic **exact** branch-and-cut (or branch-and-bound) MIP solver , e.g. Cplex or XPRESS

- **General scheme (sketch)**

1. set $h := 0$ and choose a convenient value for parameter k
2. run the MIP solver (initial upper bound = $+\infty$) for a while, until a **first feasible solution** $x^{(h)}$ is found, and let

$$S^{(h)} := \{j \in B : x_j^{(h)} = 1\}$$

be its binary support

3. add the local branching constraint $x(S^{(h)}) \geq |S^{(h)}| - k$ to the current MIP, and try to solve it exactly within a certain time limit (initial upper bound = $c^T x^{(h)}$)

4. let $x^{(h+1)}$ be the best solution found so far

5. remove the last local branching constraint $x(S^{(h)}) \geq |S^{(h)}| - k$

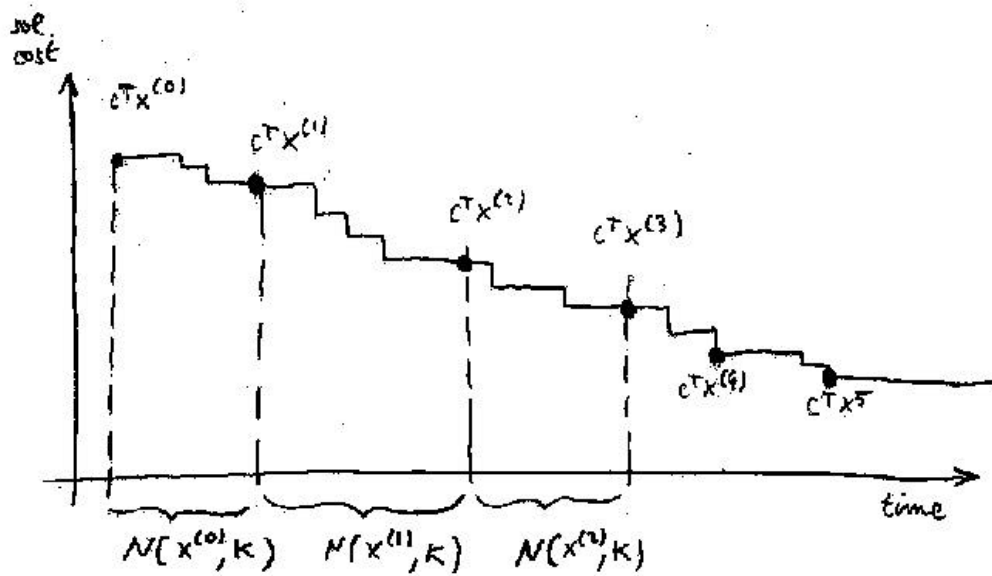
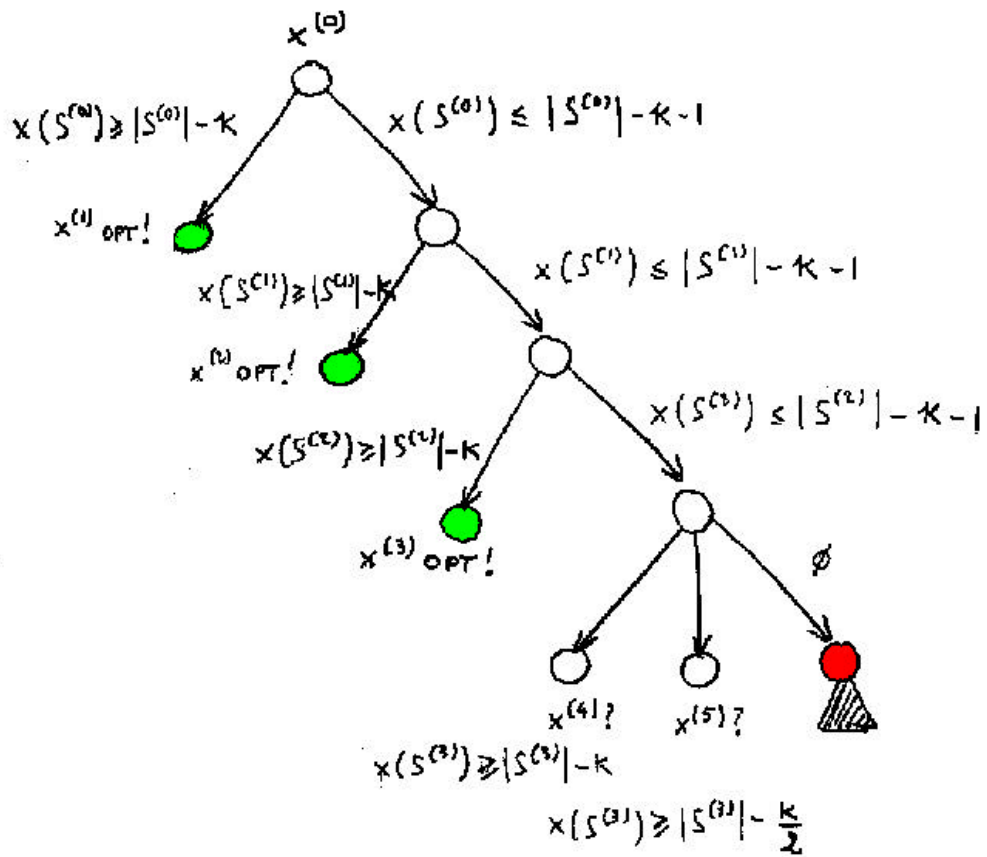
6. **GREEN flag** : if **current problem solved to proven optimality or infeasible**, then

add the “valid cut” $x(S^{(h)}) \leq |S^{(h)}| - k - 1$, set

$h := h + 1$ and repeat from step 3 (increase k if pr. infeas.)

7. if you feel lucky, reduce k and repeat from step 3

8. **RED flag** : give-up and run the MIP solver on the current problem



Example: local branching in an exact solution framework

Local branching in a heuristic solution framework

- Easy adaptation of the previous framework: when the **RED FLAG** situation occurs, use a **diversification** mechanism to find a (worse) solution $x^{(h+1)}$ to replace the current-best solution, and continue.
- Diversification by **Variable Neighbourhood Search** (Hansen & Mladenovic, 1998):

Find a solution $x^{(h+1)}$ close enough to $x^{(h)}$, but outside the current k-OPT neighbourhood, e.g.

$$x^{(h+1)} \in N(x^{(h)}, k + k/2) \setminus N(x^{(h)}, k)$$

- Run the MIP solver (initial upper bound = $+\infty$) to find the first feasible solution $x^{(h+1)}$ of the current problem amended by the **diversification constraint**

$$k + 1 \leq |S^{(h)}| - x(S^{(h)}) \leq k + k/2$$

“Akin to a random 3-OPT move after several 2-OPT moves for TSP”

Variants and extensions (under investigation)

- Tabu search by branch-and-cut:

- The black-box MIP solver is used to explore a suitable neighbourhood $N(x^H, k)$ of the current target solution x^H
- the neighbourhood is defined by a **restricted MIP model** defining a proper subset of the feasible solution space, e.g.

- not too far from the current target solution:

$$x(S^H) \geq |S^H| - K$$

- tabu moves:

$$x(S^H) \leq |S^H| - 1$$

- big-M coefficient reductions of the type

$$\sum_{j \in F} x_j \leq |F| y_j \Rightarrow \sum_{j \in F} x_j \leq \left[\sum_{j \in F} x_j^H \right] y_j$$

- possibly: problem-specific constraints ...

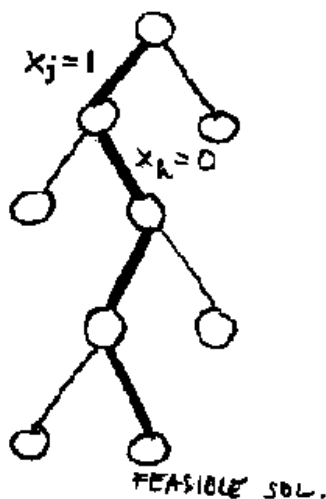
- Working with percentage gap closed

$$\left\{ \begin{array}{l} \min c^T x \\ \sum_{j \in S^H} x_j \geq |S^H| - \kappa \\ Ax \geq b \\ 0 \leq x_j \leq 1, j \in \beta \end{array} \right\} \quad \left\{ \begin{array}{l} \kappa := \left\lceil \max_{j \in S^H} \sum x_j \right\rceil \\ c^T x \leq \alpha LB + (1-\alpha) UB \\ Ax \geq b \\ 0 \leq x_j \leq 1, j \in \beta \end{array} \right.$$

"Fixed κ "

"% gap closed"

- Smart backtracking



$$F_1 := \{j \in \beta : x_j = 1\}$$

$$F_0 := \{j \in \beta : x_j = 0\}$$

$$\underbrace{\sum_{j \in F_0} x_j + \sum_{j \in F_1} (1-x_j)}_{n. \text{ of flipped vars.}} \leq \kappa$$

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Hard nesting problem (broken glass) glass4.lp

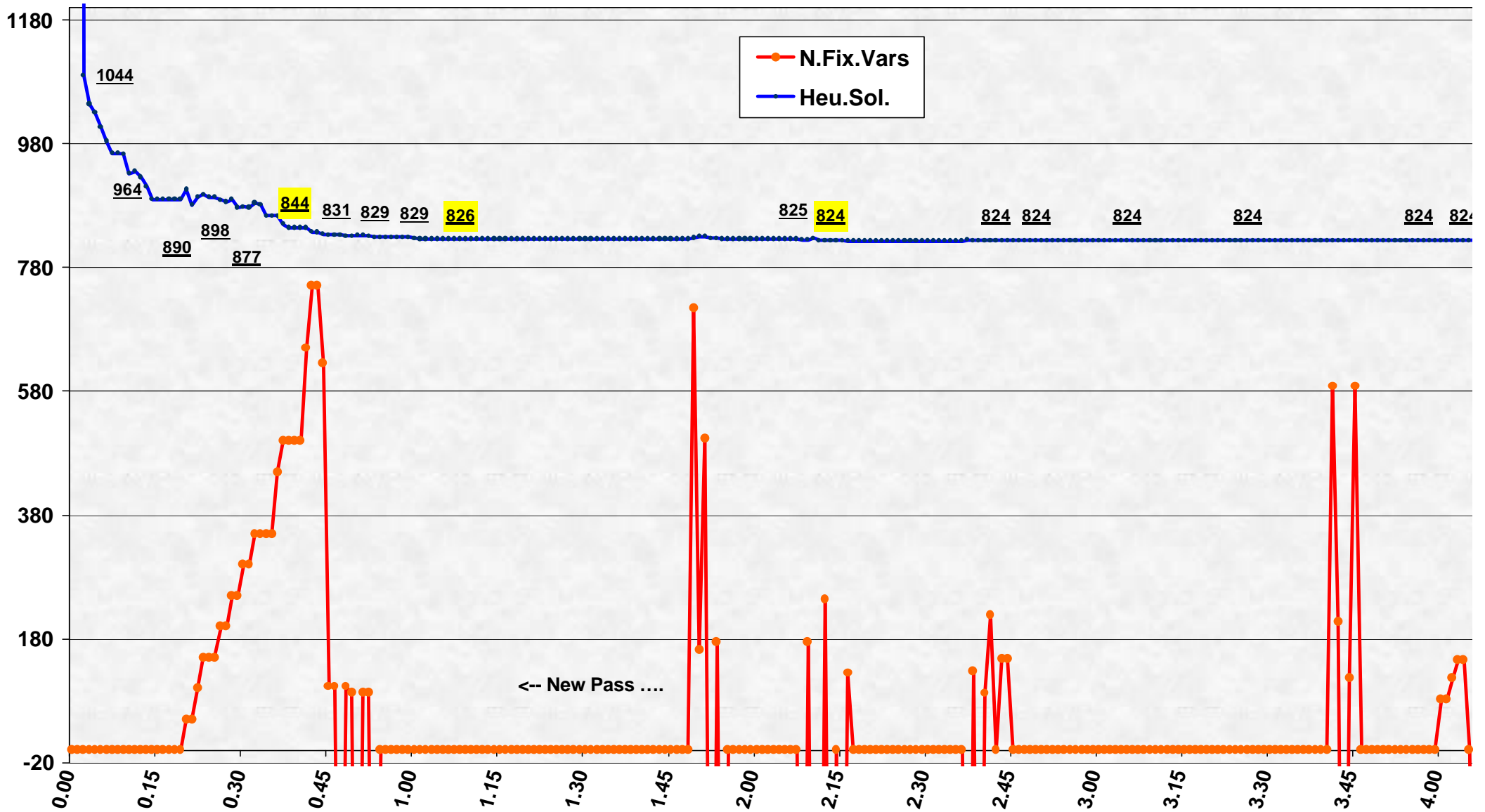
CPLEX 7.0: MIP emphasis: optimality

➤	Elapsed b&b time =	4.42 sec.	2.9334e+09
➤	Elapsed b&b time =	14.83 sec.	2.8334e+09
➤	Elapsed b&b time =	68.20 sec.	2.3000e+09
➤	Elapsed b&b time =	100.58 sec	2.2800e+09
➤	Elapsed b&b time =	354.20 sec.	2.1000e+09
➤	Elapsed b&b time =	1062.98 sec.	2.0867e+09
➤	Elapsed b&b time =	1257.15 sec.	2.0500e+09
➤	Elapsed b&b time =	3436.65 sec.	2.0125e+09
➤			1.9500e+09
➤	Elapsed b&b time =	3922.70 sec.	1.9334e+09
➤	Elapsed b&b time =	5711.65 sec.	1.8625e+09
➤	Elapsed b&b time =	6119.68 sec.	1.8500e+09
➤	Elapsed b&b time =	10770.95 sec.	1.8133e+09
➤	Elapsed b&b time =	14460.53 sec.	1.8000e+09

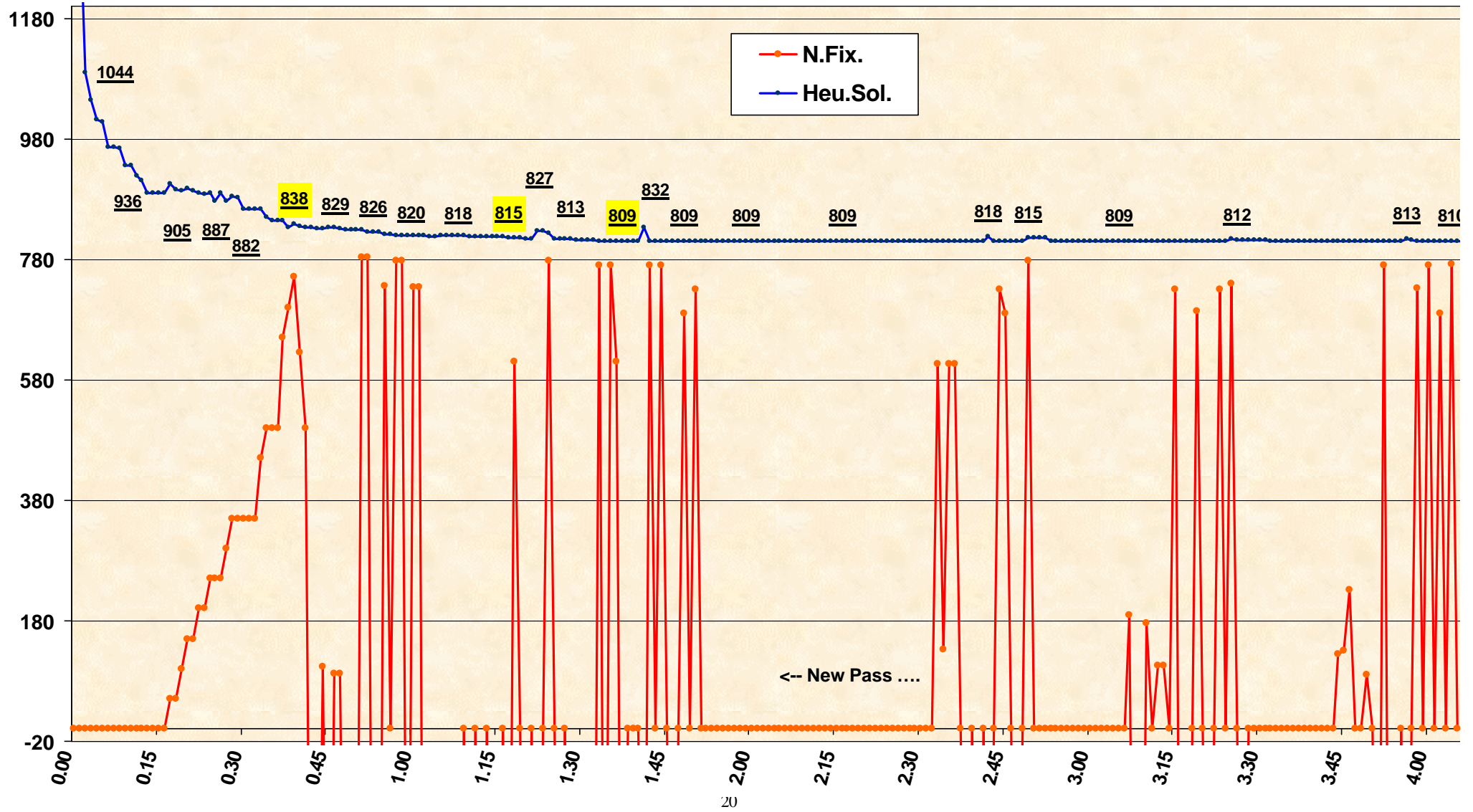
CPLEX 7.0 & Local Branching (k=10)

➤	Time =	1.9	best sol =	2,933,355,933.33
➤	Time =	302.1	best sol =	1,850,015,600.00
➤	Time =	602.3	best sol =	1,800,015,033.33
➤	Time =	902.5	best sol =	1,750,015,100.00
➤	Time =	1202.7	best sol =	1,744,458,422.22
➤	Time =	1803.0	best sol =	1,640,013,670.00
➤	Time =	7137.6	best sol =	1,600,015,950.00
➤	Time =	17513.9	best sol =	1,600,013,400.00

Example: 800+ driver duties at NSR (the Dutch railways) computed by TURNI without local branch



Example: 800+ driver duties at NSR (the Dutch railways) computed by TURNI with local branch



Hard crew scheduling (NSR): nsr8k.lp

Just the TURNI core problem of the previous real-world crew scheduling instance

CPLEX 7.0: MIP emphasis: optimality

```
Root relaxation solution time = 6509.12 sec.
```

Nodes		Cuts/				
Node	Left	Objective	IInf	Best Integer	Best Node	Gap
0	0	1.7501e+07	3653		1.7501e+07	
		1.7501e+07	3673		Fractcuts: 41	
*	0+	4.1831e+08	0	4.1831e+08	1.7501e+07	95.82%
	10	1.7507e+07	3680	4.1831e+08	1.7501e+07	95.82%
*	10+	3.2858e+08	0	3.2858e+08	1.7501e+07	94.67%
	20	1.7523e+07	3658	3.2858e+08	1.7501e+07	94.67%
	30	1.7531e+07	3618	3.2858e+08	1.7501e+07	94.67%

Final Sol. after 36,000.0 sec.s 328,581,877.40

CPLEX 7.0 & Local Branching (k=10)

➤	Time = 12,820.0	best sol = 418,308,979.25
➤	Time = 13,956.1	best sol = 123,196,426.17
➤	Time = 16,356.4	best sol = 51,215,212.14
➤	Time = 18,759.4	best sol = 42,812,004.17
➤	Time = 21,160.0	best sol = 24,064,072.04
➤	Time = 23,561.2	best sol = 23,189,634.03
➤	Time = 28,362.8	best sol = 22,812,458.03
➤	Time = 33,167.6	best sol = <u>22,757,539.03</u>

TURNI: 809-duty sol. in about 5,400 sec.s 18,257,835.61