Thinning out Steiner Trees

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The (almost) full #mozartballs family #dimacs #dimacs11





11th DIMACS Implementation Challenge in Collaboration with ICERM: **Steiner Tree Problems**

The 11th DIMACS challenge was on Steiner Tree Problems

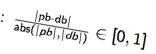
Problems Addressed The Challenge addresses the Steiner Tree problem and its variants, including

- 2. classical geometric Steiner problem (including rectilinear, Euclidear
- 3. prize-collecting Steiner trees;
- 4. dynamic Steiner trees;
- 5. node-weighted Steiner trees; 6. generalized Steiner trees (Steiner forest);
- 7. obstacle-avoiding Steiner trees;
- 8. directed Steiner trees (arborescences); rooted Steiner trees with height (or hop) constraints;
- ▶ Mean Value
 - geometric mean for primal bound

 - ▶ shifted geom. mean (shift 1.0) for primal integral and time arithmetic mean for gap
- ▶ Points-based method

Quality Challenge

- Primal Bound upon termination
- Pareto Challenge
 - Primal Integral
- ▷ Gap-based Exact Challenge
- primal-dual gap upon termination: $\frac{|pb-db|}{abs(|pb|,|db|)} \in [0,1]$
- D Time-based Exact Challenge time to solve (or time limit)





Vienna and Padua together

- ASSOCIAZIONE MOZART-ITALIA PADOVA
- A bunch of effective codes initially provided by Vienna's team
- All codes re-engineered and re-tuned to improve performance
- New exact and heuristic codes implemented (this talk)
- Initial filter to select the actual code to be run based on the instance type
- Four codes (all based on Cplex 12.6) finally submitted to DIMACS's challenge

#MozartBalls

#StayNerd #MozartDuet #HedgeKiller exact, single & multi-threaded

heuristic, single & multi-threaded heuristic, multi-threaded heuristic, multi-threaded

SPG, (R)PCSPG, MWCS, DCST SPG, PCSPG SPG, PCSPG SPG, PCSPG





Aussois, January 2015







Challenge results

Detailed final results posted by Gerald Gamrath at

http://dimacs11.cs.princeton.edu/contest/challenge-results.pdf

Many variants / scores → not a single winner...

... however #MozartBalls ranked first in many categories





Stephen J Maher @sj_maher · 5 dic #DIMACS Challenge: results are in. #mozartballs are a clear winner. Winning most of the variants they entered.



Steiner Tree Problem

Definition (The prize-collecting Steiner tree problem (PCSTP))

Given an undirected graph G = (V, E) with a (possibly empty) set of *real* terminals $T_r \subset V$, edge costs $c : E \mapsto \mathbb{R}^+$ and node revenues $p : V \mapsto \mathbb{R}^+$, the PCSTP is to find a subtree \mathcal{T} that spans all real terminals and such that the cost

$$c(\mathcal{T}) = \sum_{e \in E[\mathcal{T}]} c_e + \sum_{i \notin V[\mathcal{T}]} p_i$$

is minimized.

Definition (Potential terminals)

Among the nodes $i \in V \setminus T_r$, only those with revenues $p_i > 0$ such that at least one adjacent edge is strictly cheaper than p_i are considered as potential leaves. These nodes are referred to as *potential terminals*, and the associated set is denoted by T_p :

$$T_p = \{ v \in V \setminus T_r \mid \exists \{u, v\} \text{ s.t. } c_{uv} < p_v \}.$$

Hypercube instances

 Since the beginning of our study we focused on the (in)famous hypercube instances introduced by

Isabel Rosseti, Marcus Poggi de Aragao, Celso C. Ribeiro, Eduardo Uchoa, and Renato F. Werneck. New benchmark instances for the Steiner problem in graphs. In Extended Abstracts of the 4th Metaheuristics International Conference (MIC'2001), pages 557–561, Porto, 2001.

2.1 Hypercubes (hc)

Graphs in this series are d-dimensional hypercubes, with $d \in \{6, ..., 12\}$. For each value of d, the corresponding graph has 2^d nodes and $d \cdot 2^{d-1}$ edges. These graphs are bipartite, and both partition subsets have the same cardinality. The vertices in one of such subsets become terminals ($|X| = 2^{d-1}$).

- Real terminals only
- All edges have cost 1
- Very symmetrical very difficult LPs very hard even for heuristics

Name		E	$ \mathbf{T} $	DC	Opt
hc6u	64	192	32	NPm	39
hc7u	128	448	64	NP?	77
hc8u	256	1024	128	NP?	148
hc9u	512	2304	256	NP?	292
hc10u	1024	5120	512	??	582
hc11u	2048	11264	1024	??	1154
hc12u	4096	24576	2048	??	2275

Looking for more effective models

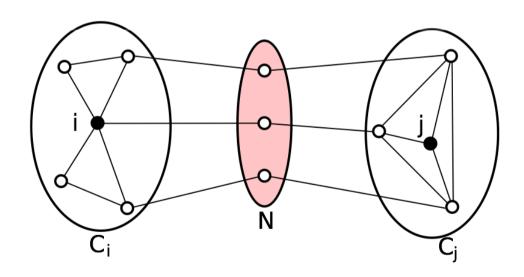
- Many hard instances (including hc*u) involve uniform edge costs
- For these instances, edge costs can easily be moved to nodes ...
 - ... hence edge variables become **redundant** (actually, **harmful** as they add a lot of symmetry and overload the model with useless cuts)
- We better work on the space of node-variables only ...
 - ... and only impose **connectivity of the subgraph** induced by the selected nodes, as in

Eduardo Alvarez-Miranda, Ivana Ljubic, and Petra Mutzel. **The maximum weight connected subgraph problem**. In *Facets of Combinatorial Optimization: Festschrift for Martin Groetschel*, 245–270, Springer, 2013.

Node separators

Definition (Node Separators)

For $i, j \in V$, a subset $N \subseteq V \setminus \{i, j\}$ is called (i, j) **node separator** iff after eliminating N from V there is no (i, j) path in G. N is minimal if $N \setminus \{i\}$ is not a (i, j) separator, for any $i \in N$. Let $\mathcal{N}(i, j)$ denote the family of all (i, j) separators.



Model for uniform edge costs

Node-based MIP model

Shift uniform edge costs *c* into node revenue:

$$\tilde{c}_{\mathsf{v}} = c - p_{\mathsf{v}}, \qquad \forall \mathsf{v} \in V$$

Let

$$T = T_r \cup T_p$$
 $P = \sum_{v \in V} p_v$

$$\min \qquad \sum_{v \in V} \tilde{c}_v y_v + (P - c) \tag{1}$$

s.t.
$$y(N) \ge y_i + y_j - 1$$
 $\forall i, j \in T, i \ne j, \forall N \in \mathcal{N}(i, j)$ (2)

$$y_{\nu} = 1 \qquad \forall \nu \in T_r \qquad (3)$$

$$y_{\nu} \in \{0,1\} \qquad \forall \nu \in V \setminus T_r \qquad (4)$$

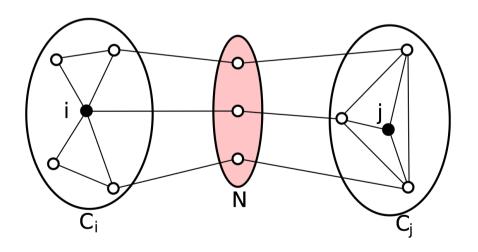
where
$$y(N) = \sum_{v \in N} y_v$$
.

Connectivity cut separation

Data: Infeasible solution defined by a vector $\tilde{y} \in \{0,1\}^n$ with $\tilde{y}_i = \tilde{y}_j = 1$, C_i being the connected components of $G_{\tilde{v}}$ containing i, and $j \notin C_i$.

Result: A minimal node separator N that violates inequality (2) with respect to i, j.

Delete all edges in $E[C_i \cup A(C_i)]$ from GFind the set R_j of nodes that can be reached from jReturn $N = A(C_i) \cap R_j$



Basic model

Node-degree inequalities:

$$y(A_i) \ge \begin{cases} y_i, & \text{if } i \in T \\ 2y_i, & \text{otherwise} \end{cases}$$
 (5)

Lifted version for potential terminals:

$$2\sum_{v\in A_i:c_{vi}< p_i}y_v+\sum_{v\in A_i:c_{vi}\geq p_i}y_v\geq 2y_i\quad\forall i\in T_p.$$

• 2-Cycle inequalities:

$$y_i \le y_j \quad i \in V, j \in T_p, c_{ij} < p_j \tag{6}$$

Branch & Cut code

- Built on top of IBM ILOG Cplex 12.6
- Initial compact relaxation → basic model (quite tight for some instances)
- Very basic implementation with connectivity cuts separated for integer points only (lazycut callback)
- Cplex's cuts at default level (but 0-1/2 cuts that are set to aggressive)
- Cplex's primal heuristics at default level (local-branching set to on)
- More elaborated implementations tried → worse performance
- Local branching heuristic using the B&C itself as a black-box solver

$$\sum_{j \in V: \tilde{y}_j = 1} y_j \ge |\{j \in V: \tilde{y}_j = 1\}| - r \tag{7}$$

- invoked a few times before attempting the exact solution
- used to gather primal information (better and better feasible solutions)
 as well as dual information (connectivity cuts) → restart policy

Exact solver: basic framework

```
Data: Input graph G, instance of the STP/PCSTP/DCST/MWCS,
       iteration and time limits.
Result: A (sub)-optimal solution Sol.
S_{init} =InitializationHeuristics()
k = 1, CutPool = \emptyset, S_{lb} = \emptyset
Choose Sol from the solution pool S_{init}.
while (k \le maxLBiter) and (time\ limit\ not\ exceeded) do
    (Sol, CutPool, S_{lb}) = LocalBranching(Sol, CutPool, S_{lb}, seed)
    Choose Sol from the solution pool S_{init}. Change seed.
    if k \mod 10 == 0 then
       Sol_{recomb} = \mathsf{Recombine}(\mathcal{S}_{\mathrm{lb}})
       if cost(Sol_{recomb}) < cost(Sol) then
        Sol = Sol_{recomb}
end
Sol = BranchAndCut(CutPool, Sol, TimeLim)
return Sol
```

Comparing exact models

Table 4. Uniform STP instances (GAPS and SP datasets). Proven optimal solutions in boldface. Previous best known solutions given in brackets. Column Time gives the computing time for proving the optimality (or, the time limit, otherwise). Columns UB and LB show upper and lower bounds obtained by the (x, y)-model, within the time limit of two hours, respectively.

				y-mo	odel		(x, y)	/)-mode	l
name	V	E	T	OPT	Time (s.)	UB	LB	gap	Time (s.)
s1	64	192	32	10	0.03	10	10	0.0%	0.01
s2	106	399	50	73	0.10	73	73	0.0%	1.67
s3	743	2947	344	514	0.29	514	508	1.19%	7200.00
s4	5202	20783	2402	3601	1.72	3601	3515	2.39%	7200.00
s5	36415	145635	16808	25210	30.06	25210	24448	3.02%	7200.00
$\overline{\text{w}13\text{c}29}$	783	2262	406	507 (508)	0.61	507	507	0.00%	167.17
w23c23	1081	3174	552	689 (694)	196.19	689	689	0.00%	1114.00

Benders-like heuristic

Steps

- Extract a relaxation of current model
- Solve relaxation heuristically
- Repair solution through local branching centered at the (infeasible) solution found
- 4 Add the connectivity cuts generated by local branching to the relaxation
- 6 Repeat



Benders-like heuristic

```
Data: Time/iteration limits.
Result: Feasible solution Sol, cut pool CutPool.
Sol = \text{dummy solution of very large cost}
CutPool = \emptyset
while (time/iteration limit not exceeded) do
   Heuristically solve a relaxation of the current model (including all cuts
   in CutPool) and let Sol^R be the possible disconnected solution found
   Add the local branching constraint (7) centered on Sol^R to the
   unrelaxed model
   (Sol', CutPool') = \texttt{BranchAndCut}(CutPool, Sol, TimeLim, SolLim)
   Remove the local branching constraint from the unrelaxed model
   if cost(Sol') < cost(Sol) then
       Sol = Sol'
   end
    CutPool = CutPool \cup CutPool'
end
return Sol
```

Set covering is everywhere!

- For pure instances (real terminals only) the node model can be interpreted as a huge **set covering problem** $y(N) \ge y_i + y_j 1 \text{ with } i, j \in T_r \Rightarrow y(N) \ge 1$
- Given the basic model + pool of connectivity cuts, the relaxation in the Benders-like framework can then be solved by a specialized set-covering heuristic → we used the CFT heuristic of

Alberto Caprara, Matteo Fischetti, and Paolo Toth. A heuristic method for the set covering problem. Operations Research, 47(5):730–743, 1999.

 For non-uniform instances, blurred version → in the "relaxation", edge costs are heuristically split among adjacent nodes (so the node model applies)



Very fast and effective, in particular, for hc*u instances



Set-covering heuristic results

Table 3. Our very preliminary results for unsolved uniform STP instances of the PUC class

name	${\bf Previous Best}$	NewBest	TURNI time (s.)	Repair time (s.)
bip52u	234	234	60.0	2.14
bip62u	220	219	60.0	0.03
bipa2u	341	338	60.0	0.05
hc9u	292	292	60.00	1.53
hc10u	581	575	60.00	4.52
hc11u	1154	1145	1800.00	0.07
hc12u	2275	2267	60.00	6.68

problem	instance	best UB	Time	#threads
STP	hc11u	$1144 \\ 2256 \\ 236158$	474	8
STP	hc12u		4817	8
STP	hc12p		4411	4
PCSTP		751	298	8
PCSTP		1492	632	16

Thanks for your attention

Full paper

M. Fischetti, M. Leitner, I. Ljubic, M. Luipersbeck, M. Monaci, M. Resch, D. Salvagnin, and M. Sinnl, **Thinning out Steiner trees: a node-based model for uniform edge costs**, Technical Report DEI, University of Padua, 2014.

and slides available at

http://www.dei.unipd.it/~fisch/papers/

http://www.dei.unipd.it/~fisch/papers/slides/

