

# An Exact Algorithm for Steiner Tree Problem on Graphs

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**Abstract:** The paper presents a new original algorithm for solving Steiner tree problem on graph. The algorithm is simple and intended for solving problems with relatively low dimensions. It is based on use of existing open source software for solving integer linear programming problems. The algorithm is tested and shown very efficient for different randomly generated problems on graphs of up to 50 nodes, up to 10 terminals and average node degree 7.

**Keywords:** Steiner tree problem on graph, branch and cut, algorithm, optimization

## 1 Introduction

The Steiner tree problem (STP) is met in telecommunication and energetic systems, VLSI technologies and in other network planning tasks. The problem is to find a minimal length tree which connects all terminal nodes of a given graph, and contains arbitrary number of other nodes. The problem is similar to the well known shortest spanning tree problem (SSTP), but unlike that it doesn't necessary contain all nodes of given graph. Also, a very important difference is that STP is much harder problem than SSTP.

The usual formulation of STP is the following: a connected undirected graph  $G = (N, E)$ , where  $N = \{1, \dots, n\}$  is a set of nodes and  $E \subseteq \{\{i, j\} \mid i \in N, j \in N, i < j\}$  denotes a set of edges, is given. A positive value (length, weight, etc.)  $c_e$  is associated to every edge  $e \in E$ . Also, a set  $T \subset N$  of so called terminal (Steiner) nodes is given.

**Definition 1.** Steiner tree for  $T$  in  $G$  is a subgraph  $St = (N', E')$ ,  $N' \subseteq N$ ,  $E' \subset E$  which satisfies one of the following statements:

1.  $T \subseteq N'$  and  $St$  is a tree;
2.  $\forall s, t \in T$  in  $St$  exists exactly one path from  $s$  to  $t$ .

The length of Steiner tree is the sum of lengths of all edges which it consists of. The Steiner tree problem is to find the shortest Steiner tree. This is a  $NP$ -hard problem and its decision problem variant belongs to  $NP$ -complete problem class [6]. The problem is presented in details in surveys: [4], [5], [9] and [11]. In papers [2], [8] and [7] the problem was solved to optimality using sophisticated branch and cut methods.

In this paper an original exact algorithm, generally based on branch and cut procedure, for solving Steiner tree problems on graph is proposed. It uses code from an available open source project which develops software for solving linear and integer linear programming problems. An intention was to formulate an algorithm which can be implemented in relatively short time and which will be able to solve STP of "reasonable" dimensions in "reasonable" time. To be more precise, the implementation of proposed algorithm lasted about 3-4 days. The program could solve problems of dimensions which are larger of the most of real life problems in, at most, several minutes.

In this paper, in the next section, a mathematical model of the Steiner tree problem and its explanation are given. In Section 3, the algorithm and the process of implementation are explained in details. In Section 4 the experiments are described and some conclusions about the algorithm behavior are made. In Section 5, the conclusion of the whole paper is given.

## 2 Mathematical model

A Steiner tree can be represented by a vector of binary variables  $x = (x_e)_{|E|}$ . Each element of the vector is assigned to one edge from the set  $E$ . The value of a variable  $x_e$  indicates whether the corresponding edge  $e$  is in the Steiner tree ( $x_e = 1$ ) or not ( $x_e = 0$ ). A mathematical model of the STP on undirected graphs has the following



**Algorithm 1** Simplified branch and cut algorithm for STP

1. Formulation of initial integer linear mathematical model:
  - (a) Goal function formulation: one variable is introduced for each edge and corresponding edge length is associated as a parameter to each variable.
  - (b) For each terminal node, one constraint of type (i) is formulated, so the terminal node is a single node on one side of the cut and the rest of the nodes are on the other side, i.e.  $\sum_{e \in \delta(M)} x_e \geq 1 \forall M \in \{\{t\} \mid t \in T\}$ .

The number of constraints after Step 1 will be  $|T|$ .
2. Solve the current mathematical model.
3. Check if the obtained solution is a Steiner tree, i.e. if there is a path between all pairs of terminal nodes. If so, the optimal solution was found in Step 2; the end of the procedure. Otherwise, next step.
4. If solution is not a Steiner tree, then it represents two or more unconnected subtrees. For every subtree add one type (i) constraint defined by cut  $\delta(M)$  where the nodes of that subtree belong to set  $M$ . Go to Step 2.

The proposed algorithm can be qualified as a simplified version of branch and cut method, i.e. a combination of branch and bound and cutting planes. The steps of the algorithm will be illustrated by an example. Suppose, we have to obtain the minimal Steiner tree for the given graph, illustrated in Figure 1, where four terminal nodes are marked with bigger circles.

The mathematical model created in Step 1 and updated in Step 4 will have one column for every edge. The constraints added in the first step are necessary to provide that every terminal is connected to, at least, one edge. In the example, four constraints will be added in the first step – one for every terminal. They will ensure that at least one edge is connected to each terminal. In Figure 2, the four cuts are marked as open curves surrounding each terminal node, and the edges marked by dashed lines are candidates to be in the first solution.

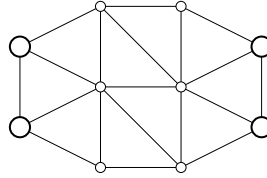


Figure 1: Initial graph

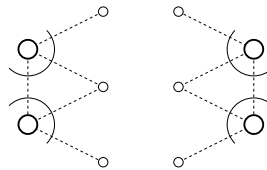


Figure 2: First step

The possible solution after Step 2 could be like one in Figure 3. As mentioned above, if the obtained solution is not a Steiner tree, it will consist of several subtrees. The number of subtrees generally can be between 2 and  $|T|$ . For determining if the obtained solution is a Steiner tree (in Step 3), Dijkstra's shortest path algorithm was used.

In Step 4 a new constraint for each subtree is added. In the example, three new constraints, corresponding to three subtrees shown in Figure 3, are added. On each image of Figure 4, one cut (represented by the curve surrounding the subtree) and edges (dashed lines) among which, at least one will be in the next solution are shown.

After the next optimization, a possible solution could be like the one shown in Figure 5. The solution satisfies all added constraints and the graph structure is a Steiner tree. Without further checking, we can claim that it is an

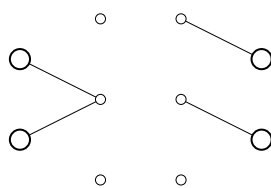


Figure 3: Possible solution after the first iteration

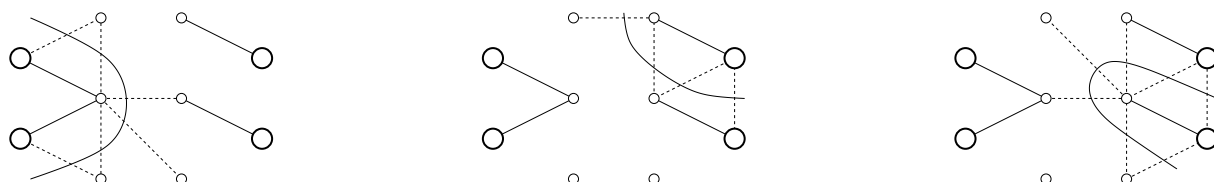


Figure 4: Three new constraints

optimal solution of the given Steiner tree problem.

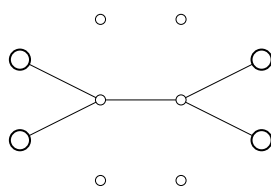


Figure 5: Final – feasible and optimal solution

The main criteria for determining data structures were access speed and simplicity of implementation. The amount of used memory was not considered because of relatively small graph dimensions of target instances. The logical choice were static structures (vectors and matrices). The realized structures enabled fast data access and mapping between graph structure (realized through neighborhood matrix) and vector of edges with all corresponding attributes.

The algorithm was implemented in C language and it is compiled and tested on Linux operation system (with gcc - GNU C Compiler) on PC with Pentium® III processor on 600 MHz and 384 MB of RAM. The implemented program supports so called *STP* format [15] of Steiner tree problems, and it is compatible with the library of standard Steiner tree problems [14].

## 4 Experimental results

The developed program was tested on different problem instances. A characteristic of the Steiner tree problem is that the complexity of a procedure for its solving depends on three attributes: number of nodes ( $n$ ), number of edges and number of terminal nodes ( $t$ ). The largest dimensions of instances successfully solved by the program were  $[n/t]$ : 20/10, 32/8 and 50/5, with the average node degree 7. The solving procedures lasted between 2 seconds and 2 minutes for the most of the instances. These dimensions may look modest in comparison with those with several thousands nodes which were successfully solved as it was reported in papers [2] and [7]. However, comparing the procedure complexity, simplicity of implementation and the fact that in many real-life telecommunication planning processes, even smaller size problems may appear, the proposed implementation could be very useful.

To get a more precise insight in the behavior of the algorithm, experiments were performed with instances where some parameters were varied. In the following table, results (mean values and standard deviation) obtained by experiments where every dimension was tested on 15 randomly generated instances are given. In the columns named “No. of rows” the number of constraints needed to obtain an optimal solution (the optimization in the last iteration) is given. The columns named “No. of iterations” represent a number of iterations of Algorithm 1 when

passing through Steps 2-4, i.e. the total number of solved binary subproblems (in Step 2). The columns “Time” represent CPU time spent to solve the problem. The column “ANCAI” shows the average number of constraints added per iteration. It’s obtained by formula:  $\frac{No.rows-t}{No.iterations}$  where  $t$  represents the number of terminal nodes, i.e. the number of constraints added in initial mathematical model in Step 1.

Table 1: The complexity analysis of the algorithm

Instance dimensions		No. of rows		No. of iterations		Time [sec.]		AN-CAI
$[n/t]$		$\bar{x}$	$\sigma$	$\bar{x}$	$\sigma$	$\bar{x}$	$\sigma$	
1		2	3	4	5	6	7	8
1	20/5	25	8.9	7.5	3.0	0.2	0.3	2.7
2	32/5	38	22.1	12.3	8.0	3.0	9.7	2.7
3	50/5	60	28.3	20.0	11.3	7.2	13.3	2.8
4	20/8	65	20.3	19.5	8.5	11.5	27.5	2.9
5	32/8	88	31.9	26.3	12.0	34.5	46.7	3.0
6	20/10	83	15.5	23.6	5.2	13.4	9.9	3.1

On the basis of data given in Table 1, some conclusions can be made. The fact that complexity of STP grows with the number of nodes is obvious from the first three rows. Although the speed of the growth cannot be determined exactly on the basis of so small sample, it is obviously nonlinear – probably exponential. The more interesting conclusion is that the growth of complexity is faster by changing the number of terminal nodes than the total number of nodes in graph. Comparing rows 1, 4 and 6, a kind of “explosion” of complexity can be seen: when the number of terminals was increased two times, execution time was increased 67 times. Similar conclusion can be made observing rows 2 and 5: addition of three terminals resulted in execution time increase of more than 11 times.

The most important analysis here concerns the number of constraints needed to get an optimal solution. According the Column 2, that growth is almost linear in problem size. We cannot be certain if it is linear, but it is definitely not exponential. Finally, we can conclude that, although the number of constraints in model (1) grows exponentially, the number of constraints necessary to obtain an optimal solution grows much slower. The number of iterations grows even less. The parameter in Column 8 is also interesting. The average number of constraints added in each iteration also represents the average number of subtrees obtained in each sub solution. It seems that the relatively small values in Column 8 doesn’t depend much on the number of terminal nodes. The explanation could be that the current solution of the solving procedure relatively quickly forms a structure which consists of a few subtrees each containing several terminals. This may contradict to the previous statement that complexity depends more on the number of terminals than on the total number of nodes, because, what influences the number of iterations is the number of unconnected subtrees, and not the number of terminals. One possible explanation is that the structure of subtrees changes, so subtrees contain different terminals in different iterations. It is also important to have in mind that every iteration lasts more than the previous one because every mathematical model has more rows (constraints) than the previous one.

Yet another interesting thing from Table 1 is relatively big dispersion (represented by standard deviation) of data obtained by different randomly generated instances with same characteristics. This is a consequence of a nature of the algorithm (which is nondeterministic polynomial). It is impossible to predict the number of iterations necessary to obtain final solution. In the worst case it can be exponential.

## 5 Conclusion

The first impressions and experiment conclusions indicate that the proposed algorithm can be efficiently used when there is a need for rapid development of an algorithm for solving smaller size Steiner tree problems. Although the worst case number of constraints needed to obtain a final solution remains exponential, the algorithm have shown a kind of “good behavior” – in all solved examples the number stayed relatively low. Although the exponential complexity of the branch and bound method (Step 2 of the algorithm) remains, instances with acceptable dimensions can be solved in real time.

The idea of successive adding violated constraints could be also applied to some other problems whose mathematical models have an exponential number of constraints. Concerning that, some new researches have been planned, where the idea would be applied to the traveling salesman problem (TSP). Namely, the so called DFJ formulation of TSP [3] also has exponential number of constraints, but it has shown a good behavior in relaxation based algorithms.

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