

# On the Largest Common Subgraph Problem

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March 13, 1990

## Abstract

The largest common subgraph problem (LCSG, for short) asks to find a common connected subgraph of the given two graphs  $G_1$  and  $G_2$ , with the largest number of edges. In this paper, we develop polynomial time algorithms for LCSG when both  $G_1$  and  $G_2$  are trees.

## 1 Introduction

Given two graphs  $G_1$  and  $G_2$ , *the largest common subgraph problem* (LCSG, for short) asks to find a common connected subgraph of both  $G_1$  and  $G_2$ , with the largest number of edges. These problems appear in detection and recognition of the largest connected substructure possessed commonly by a plural number of chemical structures in the structure-activity studies, where exploring functional groups or particular structural fragments common to organic molecules which have the same biological or pharmacological function is one of the significant issues. Several computational methods for the problem of finding largest common substructures in two or more molecules were suggested (see e.g., [2, 4, 9, 10]). Generally, however, their algorithms arisen in such application are quite time-consuming as they are performed in exhaustive manner

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based on atom-by-atom (i.e., vertex-by-vertex) comparisons. It is true that chemical structure diagrams are closely related to the graphs in their representation. Thus more efficient algorithms are now actually required for the LCSG problem in systematization of the computer-assisted design in chemistry.

LCSG in general graphs is obviously *NP*-complete as Hamiltonian circuit problem, which is known to be *NP*-complete [1], can easily be reduced to this problem where  $G_1 = C_{|V|}$  (a cycle on  $|V|$  vertices) and  $G_2 = G(= (V, E))$ . Efficient algorithms may, however, exist for some special cases.

In this paper, we develop polynomial time algorithms for LCSG when both  $G_1$  and  $G_2$  are trees.

## 2 LCSG Over a Rooted Tree

In this section, a polynomial time algorithm for LCSG is developed when both  $G_1$  and  $G_2$  are rooted trees  $(T_1, r_1)$  and  $(T_2, r_2)$ , where  $r_1$  ( $r_2$ ) is the root of  $T_1$  ( $T_2$ ). For each vertex  $v$  in  $(T, r)$ , let  $(T(v), v)$  denote the subtree of  $(T, r)$  whose root is  $v$  and spanned by all the descendants of  $v$ . We first introduce *procedure LCS* for obtaining the largest common subtree  $(T, r)$  of two rooted trees  $(T_1, r_1)$  and  $(T_2, r_2)$  where  $r$  corresponds to both  $r_1$  and  $r_2$ , respectively.

*Procedure*  $LCS((T_1, r_1), (T_2, r_2), N((T_1, r_1), (T_2, r_2)))$

*Input:* Rooted trees  $(T_1, r_1)$  and  $(T_2, r_2)$ .

*Output:* The number  $N((T_1, r_1), (T_2, r_2))$  of edges of the largest common subtree  $(T, r)$  of  $(T_1, r_1)$  and  $(T_2, r_2)$  where  $r$  corresponds to both  $r_1$  and  $r_2$ , respectively.

*begin* (Initialization)

For each leaf  $l_i$  of  $(T_1, r_1)$  and  $l'_j$  of  $(T_2, r_2)$ ,

$$N((T_1(l_i), l_i), (T_2(l'_j), l'_j)) \leftarrow 0.$$

*end*

*begin*

Let  $v_1, \dots, v_d$  and  $w_1, \dots, w_{d'}$  be sons of the roots  $r_1$  and  $r_2$ , respectively. Recursively call procedure  $LCS((T_1, r_1), (T_2, r_2), N((T_1(v_i), v_i), (T_2(w_j),$

$w_j)))$  for  $i = 1, \dots, d$ ,  $j = 1, \dots, d'$ , and construct a bipartite graph  $G = (V_1, V_2, E, c)$ , where  $V_1 = \{(T_1(v_i), v_i) | i = 1, \dots, d\}$ ,  $V_2 = \{(T_2(w_j), w_j) | j = 1, \dots, d'\}$ ,  $E = \{((T_1(v_i), v_i), (T_2(w_j), w_j)) | i = 1, \dots, d, j = 1, \dots, d'\}$  and the weight  $c((T_1(v_i), v_i), (T_2(w_j), w_j))$  of each edge  $((T_1(v_i), v_i), (T_2(w_j), w_j))$  is

$$N((T_1(w_i), w_i), (T_2(v_j), v_j)) + 1 \quad (i = 1, \dots, d, j = 1, \dots, d').$$

$N((T_1, r_1), (T_2, r_2)) \leftarrow$  the value of the maximum weight matching [7] of  $G$ .

end  $\square$

*Note 2.1.* Throughout this paper, we introduce algorithms for obtaining the number of edges of the largest common subtree. It is easy, however, to modify such algorithms so that the actual largest common subtree is obtained, although we omit the details.  $\square$

*Lemma 2.1.* Procedure LCS correctly finds the number  $N((T_1, r_1), (T_2, r_2))$  of the largest common subtree  $(T, r)$  of  $(T_1, r_1)$  and  $(T_2, r_2)$ , here  $r$  corresponds to both  $r_1$  and  $r_2$ , in  $O(n_1^2 n_2)$  time, where  $n_1$  ( $n_2$ ) is the number of vertices in  $(T_1, r_1)$  ( $(T_2, r_2)$ ).

(Proof) We shall prove this lemma by induction on  $h$ , where  $h$  is the height of the tree which is heigher than or equal to the other among  $(T_1, r_1)$  and  $(T_2, r_2)$ . This lemma is trivial when  $h = 1$ . Soppose that this lemma holds when  $h \leq H - 1$ , i.e.,  $N((T_1(v_i), v_i), (T_2(w_j), w_j))$  obtained by procedure LCS is the number of the largest common subtree of  $(T_1(v_i), v_i)$  and  $(T_2(w_j), w_j)$  for each pair  $(T_1(v_i), v_i)$  and  $(T_2(w_j), w_j)$ , where  $v_1, \dots, v_d$  and  $w_1, \dots, w_{d'}$  are sons of the roots  $r_1$  and  $r_2$ , respectively. Without loss of generality, we assume that the height of  $(T_1, r_1) \leq$  the height of  $(T_2, r_2)$ , i.e., the height of  $(T_2, r_2)$  is  $H$ . Let  $(T, r)$  be the largest common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$  where  $r$  corresponds to both  $r_1$  and  $r_2$ , and let  $(T', r_1)$  ( $(T'', r_2)$ ) be the subtree in  $(T_1, r_1)$  ( $(T_2, r_2)$ ) isomorphic to  $(T, r)$ . We consider a bipartite graph  $G = (V_1, V_2, E, c)$ , where  $V_1 = \{(T_1(v_i), v_i) | i = 1, \dots, d\}$ ,  $V_2 = \{(T_2(w_j), w_j) | j = 1, \dots, d'\}$ ,  $E = \{((T_1(v_i), v_i), (T_2(w_j), w_j)) | i = 1, \dots, d, j = 1, \dots, d'\}$  and the weight  $c((T_1(v_i), v_i), (T_2(w_j), w_j))$  of each edge  $((T_1(v_i), v_i), (T_2(w_j), w_j))$  is

$$N((T_1(v_i), v_i), (T_2(w_j), w_j)) + 1 \quad (i = 1, \dots, d, j = 1, \dots, d').$$

Consider the set of edges  $M = \{((T_1(v_i), v_i), (T_2(w_j), w_j)) | (T_1(v_i), v_i) \text{ and } (T_2(w_j), w_j) \text{ corresponds to the same subtree of } (T, r)\} \subset E$ . Then  $M$

must be the maximum weight matching of  $G$  as, otherwise, let  $M'$  be a matching of  $G$  whose value is greater than that of  $M$  and we can construct a larger common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$  from  $M' \subset E$ , where  $r$  corresponds to  $r_1$  and  $r_2$ . This, however, contradicts the fact that  $(T, r)$  is the largest common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$ , proving that the common subtree obtained by the algorithm is also the largest one when  $h = H$ , as desired. Thus we have proved this lemma by induction.

The time complexity is now analyzed. Note that the most time-consuming part is the maximum weight matching. We assume here that  $h_1 \leq h_2$  where  $h_1$  ( $h_2$ ) is the height of  $T_1$  ( $T_2$ ). The maximum weight matching is solved at vertices within distance  $h_1$  from  $r_2$  in  $T_2$  and at each vertex  $v$  whose degree is  $k_2$  matched with vertex in  $T_1$  whose degree is  $k_1$ , the maximum weight matching is solved in  $O(k_1^2 k_2)$  time by a well-known primal-dual type algorithm (see e.g., [7]). Thus the time complexity in total is

$$\sum O(k_1^2 k_2) = O(n_1^2 n_2). \quad \square$$

Now we turn to LCSG where the root of common subtree may correspond to any vertex in  $(T_1, r_1)$  ( $(T_2, r_2)$ ).

*Note 2.2.* The root  $r$  of the largest common subtree  $(T, r)$  must correspond to at least one of  $r_1$  (of  $(T_1, r_1)$ ) or  $r_2$  (of  $(T_2, r_2)$ ) as:

Let  $(T', v_1)$  ( $(T'', v_2)$ ) be the subtree isomorphic to  $(T, r)$  in  $(T_1, r_1)$  ( $(T_2, r_2)$ ) and consider path  $s_1$  in  $(T_1, r_1)$  ( $s_2$  in  $(T_2, r_2)$ ) from  $v_1$  to  $r_1$  (from  $v_2$  to  $r_2$ ) and, without loss of generality, we assume that  $s_1 \leq s_2$  holds. Let  $s'$  be the path in  $(T_2, r_2)$  from  $v_2$  to an ancestor  $v'$  of  $v_2$ , whose length is  $s_1$ . Then by appending  $s_1$  ( $s'$ ) to  $(T', v_1)$  ( $(T'', v_2)$ ), we have a larger common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$ , contradicting the assumption.  $\square$

Based on Note 2.2, we have the following algorithm LCRT for obtaining the largest common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$ .

*Algorithm LCRT* $((T_1, r_1), (T_2, r_2), N)$

*Input:* Rooted trees  $(T_1, r_1)$  and  $(T_2, r_2)$ .

*Output:* The number  $N$  of edges of the largest common subtree  $(T, r)$  of  $(T_1, r_1)$  and  $(T_2, r_2)$ .

*Step 1.* For each vertex  $v$  in  $(T_1, r_1)$  call  $\text{LCS}((T_1(v), v), (T_2, r_2), N((T_1(v), v), (T_2, r_2)))$ .

*Step 2.* For each vertex  $v$  in  $(T_2, r_2)$  call  $\text{LCS}((T_2(v), v), (T_1, r_1), N((T_2(v), v), (T_1, r_1)))$ .

*Step 3.*

$$N \leftarrow \max \left\{ \max_{v \text{ in } (T_1, r_1)} N((T_1(v), v), (T_2, r_2)), \max_{v \text{ in } (T_2, r_2)} N((T_2(v), v), (T_1, r_1)) \right\}$$

*Step 4.* Halt.  $\square$

*Theorem 2.1.* Algorithm LCRT solves LCSG over rooted trees in  $O(n_1^2 n_2^2)$  time, where  $n_1$  ( $n_2$ ) is the number of vertices in  $(T_1, r_1)$  ( $(T_2, r_2)$ ).

(Proof) The correctness is based on Note 2.2 and that of procedure LCS, hence that of Lemma 2.1.

Step 1 in algorithm LCRT is performed in  $O(n_1 n_1 n_2^2) = O(n_1^2 n_2^2)$  time and Step 2 of algorithm LCRT is executed in  $O(n_2 n_2 n_1^2) = O(n_1^2 n_2^2)$  time by Lemma 2.1. Thus the time complexity of this algorithm in total is  $O(n_1^2 n_2^2)$ .  $\square$

### 3 LCSG over an Undirectred Tree

In this section, we develop a polynomial time algorithm for the largest common subtree  $T$  of undirected trees  $T_1$  and  $T_2$ . Let  $(T_1, r_1)$  ( $(T_2, r_2)$ ) be a rooted tree obtained from  $T_1$  ( $T_2$ ) by choosing an arbitrary vertex  $r_1$  in  $T_1$  ( $r_2$  in  $T_2$ ) as a root. Then a rooted tree  $(T, r)$  isomorphic to rooted subtrees  $(T', v')$  in  $(T_1, r_1)$  and  $(T'', v'')$  in  $(T_2, r_2)$ , both corresponding to  $T$ , are also the largest common subgraphs of  $(T_1, r_1)$  and  $(T_2, r_2)$ . Conversely, let  $(T, v)$  be the largest common subtree of  $(T_1, r_1)$  and  $(T_2, r_2)$  and  $T'$  be the corresponding undirected tree obtained from  $T$  by neglecting the direction of edges and by not specifying any vertex as a root. Then  $T'$  is also the largest common subtree of  $T_1$  and  $T_2$ . Based on this observation, we have the following algorithm LCUT whose time complexity is  $O(n_1^2 n_2^2)$  where  $n_1$  ( $n_2$ ) is the number of vertices in  $T_1$  ( $T_2$ ).

*Algorithm LCUT*

*Input:* Undirected trees  $T_1 = (V_1, E_1)$  and  $T_2 = (V_2, E_2)$ .

*Output:* The number  $N$  of the edges of the largest common subtree of undirected trees  $T_1 = (V_1, E_1)$  and  $T_2 = (V_2, E_2)$ .

*Step 1.* Choose arbitrary vertices  $r_1$  of  $T_1$  and  $r_2$  of  $T_2$ , respectively, and construct rooted trees  $(T_1, r_1)$  and  $(T_2, r_2)$ . Execute algorithm LCRT( $(T_1, r_1), (T_2, r_2), N$ ).

*Step 2.* Halt.  $\square$

## 4 Concluding Remarks

We can easily apply algorithm LCUT developed in this paper to the recognition of the largest common substructure of two chemical compounds with tree structures by specifying which vertex (edge) of  $T_1$  may correspond to which vertex (edge) of  $T_2$  according to the definition of structural similarity (see e.g., [10]) which is most suitable for the purpose of each research. Although applicability of the present algorithm is limited to acyclic molecules, it would be possible to apply it to wider class of molecules which have isolated simple rings by abstracting or devising the graph representation of their structures.

Finding some other special cases in which LCSG can be solved efficiently, providing good heuristic algorithms for general graphs and developing algorithms to find the largest common subgraph of more than two graphs (see e.g., [9, 10]) seem to deserve further research. The algorithms developed in this paper may be useful for these purposes.

## References

- [1] A. V. Aho, J. E. Hopcroft and J. D. Ullman, *The Design and Analysis of Computer Algorithms*, Addison-Wesley, Reading, Mass. (1974).
- [2] J. E. Armitage, J. E. Crowe, P. N. Evans, M. F. Lynch and J. A. McGuirk, "Documentation of Chemical Reactions by Computer Analysis of Structural Changes", *J. Chem. Doc.*, 7, 209-215 (1967).
- [3] C. Berge, *Graphs and Hypergraphs*, North-Holland, Amsterdam (1973).
- [4] M. M. Cone, R. Venkataraghavan and F. W. McLafferty, "Molecular Structure Comparison Program for the identification of Maximal Common Substructures", *J. Am. Chem. Soc.* 99, 7668-7671 (1977).

- [5] M. R. Garey and D. S. Johnson, *Computers and Intractability: A Guide to the Theory of NP-Completeness*, W. H. Freeman and Company, San Francisco (1979).
- [6] F. Harary, *Graph Theory*, Addison-Wesley, Reading, Mass. (1969).
- [7] E. L. Lawler, *Combinatorial Optimization: Networks and Matroids*, Holt Rinehart and Winston, New York (1976).
- [8] S. Masuyama, "On the minimum cost subgraph problem", *Proceedings of the International Workshop on Discrete Algorithms and Complexity*, 101-106 (1989).
- [9] Y. Takahashi, Y. Sato, H. Suzuki and S. Sasaki, "Recognition of largest common structural fragment among a variety of chemical structures", *Analytical Sciences*, 3, 23-28 (1987).
- [10] T. H. Varkony, Y. Shiloach and D. H. Smith, "Computer Assisted Examination of Chemical Compounds for Structural Similarities", *J. Chem. Inf. Comput. Sci.* 19, 104-111 (1979)