

# Counting Edge Covers Sets for Estimating the Relevance of Communication Lines

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**Abstract.** Counting the number of edge covers on graphs is a #P-complete problem. We design efficient exact methods for computing the number of edge covers for acyclic graphs. Even more, we show that if a graph  $G$  does not contain intersecting cycles (any pair of cycles has not common edges) then its number of edge covers can be computed in linear time over the size of the graph  $G$ . This determines a border between efficient counting and exponential time procedures for counting the number of edge covers.

We also show how to apply the computing of the number of edge cover for estimating the relevance of the lines in a communication network, which is an important problem in the reliability analysis of a network.

**Keywords:** Counting Edge Covers, Combinatorial Algorithms, Reliability Analysis Network.

## 1 Introduction

Counting problems are not only mathematically interesting, but they arise in many applications. For example, if we want to know the probability that a propositional formula is true, or the probability that a graph remains connected given a probability of failure of an edge, we have to count to approximate such probabilities. Counting problems also arise naturally in Artificial Intelligence research. For example, some methods used in reasoning, such as computing 'degree of belief' and 'Bayesian belief networks' are computationally equivalent to counting the number of satisfying assignments to a propositional formula [8, 10]

Counting has become an important area in mathematics as well as in theoretical computer science, although it has received less attention than decision problems. Actually, there are few counting problems in graph theory that can be solved exactly in polynomial time, indeed an important line of research is to determine the class of graphs (or the class of restrictions) for which a counting problem could be solved in polynomial time.

An *edge cover* set of a graph  $G$  is a subset of edges covering all nodes of  $G$ . The problem of counting the number of edge cover sets of a graph, denoted as #Edge\_Covers, is a #P-complete problem via the reduction from #Twice-SAT to #Edge\_Covers [1].

Although the computation of #Edge\_Covers is a hard problem, it is relevant to recognize the class of instances where this problem becomes an easy problem,

that means, to identify the class of graphs where counting the number of its edge covers can be done in polynomial time. There is a scarce literature about the design of procedures for computing edge covers, and as far as we know, it is not known which is the largest polynomial class of graphs for the  $\#Edge\_Covers$  problem.

We address the computation of  $\#Edge\_Covers$  based on the topological structure of the graph. We focus on determining the topology of the graph  $G$  which allows to count the number of edge covers of  $G$  in an exactly and efficiently way. We show here that the  $\#Edge\_Covers$  problem can be computed in polynomial time for any acyclic graph.

We show the relevance to compute the number of edge cover for computing the relevance of the lines into a communication network. We show that the computation of the number of edge covers is an important value for estimating the 'strategic' value of each line of a network.

## 2 Preliminaries

A connected graph is a graph such that there exists a path between all pairs of vertices. If the graph is a directed graph, and there exists a path from each vertex to every other vertex, then it is a strongly connected graph.

A vertex cover of a graph  $G = (V, E)$  is a subset  $U \subseteq V$  that covers every edge of  $G$ ; that is, every edge has at least one endpoint in  $U$ .

An edge cover,  $\mathcal{E}$ , for a connected graph  $G = (V, E)$  is a subset of edges  $\mathcal{E} \subseteq E$  which contains edges covering all vertex of  $G$ , that is, for each  $u \in V$  there is a  $v \in V$  such that  $e = \{u, v\} \in \mathcal{E}$ .

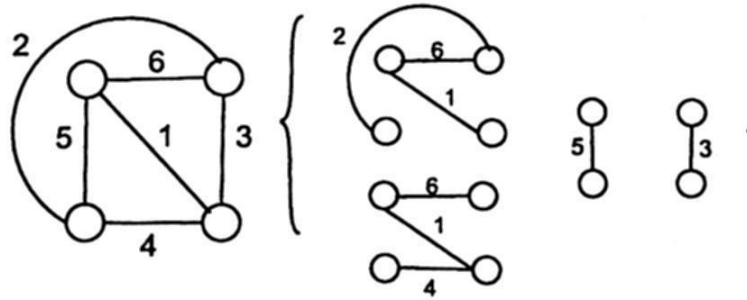


Fig. 1. Cases of edge covers

Given a connected graph  $G = (V, E)$ , let  $C_\epsilon(G) = \{\epsilon \subseteq E : \epsilon \text{ is an edge cover of } G\}$  be the set of edge-covers sets that a graph  $G$  has. Let  $NE(G) = |C_\epsilon(G)|$  be the number of different edge-covers in a graph, and given any graph  $G$ , we denote the problem of computing the number  $NE(G)$  as the  $\#Edge\_Covers$  problem.

The  $\#Edge\_Covers$  Counting Problem consists in given a connected graph, find the total number of edge covers of  $G$ . For example, in the figure 1 we show three different edge covers sets of a graph.

### 3 CEC: Counting Edge Covers

The value  $NE(G)$  for any graph  $G$ , including the case when  $G$  is a disconnected graph, is obtained as:  $NE(G) = \prod_{i=1}^k NE(G_i)$ , where  $G_i, i = 1, \dots, k$  is the set of connected components of  $G$ . We should first determine the set of connected components of  $G$ , and this procedure can be done in linear time. Then, the time complexity of computing  $NE(G)$  depends on the maximum time complexity of its connected components. Thus, we would consider just the different kinds of connected components in  $G$ , so from now on, when we mention a graph  $G$  we suppose that it consists of just one connected component.

We call *fixed edges* to the edges of a graph  $G$  that appear in all edge cover set of  $G$ . When an edge cover  $\mathcal{E}$  of a graph is being built, we distinguish between two different states of a node; we say that a node  $u$  is *free* when it has not still been covered by any edge of  $\mathcal{E}$ , while if the node has already been covered we say that the node is *cover*.

In order to present the basic procedures for counting edge covers, we consider first the case when the graph is an acyclic graph, and we start analyzing the most simple topology of an acyclic graph. An initial procedure for counting edge covers is to apply a depth first search over the input graph. A basic a recursive procedure scheme for the depth-first search, is the following.

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**Algorithm 1** Procedure  $dfs(G, v)$ 


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Mark  $v$  as discovered
for each vertex  $w \in N(v)$  do
  if ( $w$  is undiscovered) then
     $dfs(G, w)$ 
  end if
Mark  $v$  as finished
end for

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#### Case A: Counting Edge Covers on Paths

Let  $P_n = G = (V, E)$  be a path graph. We assume an order between vertices and edges in  $P_n$ , i.e. let  $V = \{v_0, v_1, \dots, v_n\}$  be the set of  $n + 1$  vertices and let  $e_i = \{v_{i-1}, v_i\}$ ,  $1 \leq i \leq n$  be the  $n$  edges of  $P_n$ .

Let  $G_i = (V_i, E_i)$ ,  $i = 0, \dots, n$  be the subgraphs induced by the first  $i$  nodes of  $V$ , i.e.  $G_0 = (\{v_0\}, \emptyset)$ ,  $G_1 = (\{v_0, v_1\}, \{e_1\})$ ,  $G_2 = (\{v_0, v_1, v_2\}, \{e_1, e_2\})$ ,  $\dots$ ,  $G_n = P_n = (V, E)$ .  $G_i, i = 0, \dots, n$  is the family of induced subgraphs of  $G$  formed by the first  $i$  nodes of  $V$ . Let  $\mathcal{CE}(G_i) = \{\mathcal{E} \subseteq E_i : \mathcal{E} \text{ is an edge cover of } G_i\}$  be the set of edge covers of each subgraph  $G_i$ ,  $i = 0, \dots, n$ .

We want to count the edge cover sets of  $P_n$  by considering the different ways of covering each node on the path. We travel by  $P_n$  in depth-first search and when a node is being visited, we consider the different ways to cover it. A path  $P_n$  has two special edges:  $e_1$  and  $e_n$  which are fixed edges and for  $i = 1, \dots, n-1$ ,  $\delta(v_i) = 2$ ,  $e_{i-1}$  and  $e_i$  are the two incident edges of  $v_i$ .

We associate with each edge  $e_i, i = 1, \dots, n$  in the path, an ordered pair:  $(\alpha_i, \beta_i)$  of integer numbers where  $\alpha_i$  expresses the number of edge cover sets in  $\mathcal{CE}(G_i)$  where the edge  $e_i$  appears in order to cover the node  $v_{i-1}$ , while  $\beta_i$  conveys the number of edge cover sets in  $\mathcal{CE}(G_i)$  where the edge  $e_i$  does not appear, since  $v_{i-1}$  has been already covered (perhaps by the edge  $e_{i-1}$ ).

Traversing by  $P_n$  in depth-first search, each pair  $(\alpha_i, \beta_i)$  is computed in accordance with the type of edge  $e_i, i = 1, \dots, n$  which is being visited. At the end of the search, the last pair  $(\alpha_n, \beta_n)$  is computed, and such pair gives the value for  $NE(P_n) = \alpha_n + \beta_n$ .

The pair  $(1, 0)$  is assigned to  $(\alpha_1, \beta_1)$  since the edge  $e_1$  is a fixed edge and  $e_1$  has to appear in all edge cover of  $P_n$ . In general, any fixed edge  $e_p$  which starts a series  $(\alpha_p, \beta_p)$  has  $(1, 0)$  as initial pair.

If we know the pair  $(\alpha_{i-1}, \beta_{i-1})$  for any  $i < n$ , then we know the number of times where the edge  $e_{i-1}$  appears or does not appear into the set of edge covers of  $G_{i-1}$ . When the edge  $e_i$  is being visited, the vertex  $v_{i-1}$  has to be covered considering for this its two incident edges:  $e_{i-1}$  and  $e_i$ . Any edge cover of  $\mathcal{CE}(G_{i-1})$  containing the edge  $e_{i-1}$  ( $\alpha_{i-1}$  cases) has already covered  $v_{i-1}$  and then the occurrence of  $e_i$  is optional. But for the edge covers where  $e_{i-1}$  does not appear ( $\beta_{i-1}$  cases) the edge  $e_i$  must appear in order to cover to  $v_{i-1}$ . Then, the number of edge covers where  $e_i$  appears is  $\alpha_{i-1} + \beta_{i-1}$  and just in  $\alpha_{i-1}$  edge covers the edge  $e_i$  does not appear. Thus, we obtain the new pair  $(\alpha_i, \beta_i)$  associated with the edge  $e_i$ , by applying the Fibonacci recurrence relation.

$$\alpha_i = \alpha_{i-1} + \beta_{i-1}; \quad \beta_i = \alpha_{i-1} \quad (1)$$

When the search arrives to the last edge  $e_n$  of the path, we have obtained the pair  $(\alpha_{n-1}, \beta_{n-1})$  and since  $e_n$  is a fixed edge, then it has to appear in all edge covers of  $P_n$ , that means,  $\alpha_n = \alpha_{n-1} + \beta_{n-1}$  and since  $e_n$  has not chance of not appearing in any edge cover of  $P_n$  then  $\beta_n = 0$ . We call *recurrence for processing fixed edges* to the recurrence:

$$\alpha_i = \alpha_{i-1} + \beta_{i-1}; \quad \beta_i = 0 \quad (2)$$

In this way, the pair associated with the last edge on a path is:  $(\alpha_n, \beta_n) = (\alpha_{n-1} + \beta_{n-1}, 0)$ . The series  $(\alpha_i, \beta_i), i = 1, \dots, n$  built based on the recurrences (1) and (2) allows to compute  $NE(P_n)$  in linear time over the number of edges in  $P_n$  in accordance with the traversing of the path in depth-first search.

We call a *computing thread* or just a *thread* to a series of pairs  $(\alpha_i, \beta_i), i = 1, \dots, n$  used for computing the number of edge covers on a trajectory of  $n$  distinct and adjacent edges. The computation of each pair  $(\alpha_i, \beta_i), i = 1, \dots, n$  is done in accordance with the type of edge  $e_i$  on the trajectory. Notice that a trajectory could be a subgraph of a more complex graph.

In a path  $P_n$ , all nodes have degree 2 or 1. But, when the current edge  $e_i$  as well as the previous one  $e_{i-1}$  considered in a computing thread are incident to a node  $v_i$  which has been already covered or for which  $\delta(v_i) > 2$ , then the edge  $e_i$  can be taken into account or not since  $v_i$  has been covered previously, and then the pair  $(\alpha_i, \beta_i)$  is computed from  $(\alpha_{i-1}, \beta_{i-1})$  by the recurrence relation:

$$\alpha_i = \alpha_{i-1} + \beta_{i-1}; \quad \beta_i = \alpha_{i-1} + \beta_{i-1} \quad (3)$$

In the following examples, we denote with  $\rightarrow$  the application of recurrence (1), with  $\xrightarrow{o}$  the application of recurrence (3) and with  $\mapsto$  the processing of fixed edges - recurrence (2).

The sequence: 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, ..., in which each number is the sum of the preceding two, is denoted as the Fibonacci series. The numbers in the sequence, known as the Fibonacci numbers, will be denoted by  $F_i$  and we formally define them as:  $F_0 = 0$ ;  $F_1 = 1$ ;  $F_{i+2} = F_{i+1} + F_i$ ,  $i \geq 0$ . Each Fibonacci number can be bounded from above and from below by  $\phi^{i-2} \geq F_i \geq \phi^{i-1}$ ,  $i \geq 1$ , where  $\phi = \frac{1}{2} \cdot (1 + \sqrt{5})$  is known as the 'golden ratio'.

**Example 1** Let us consider the path  $P_5$  (see figure 2). The computing thread for  $P_5$  is:  $(\alpha_i, \beta_i)$ ,  $i = 1, \dots, 5$ :  $(1, 0) \rightarrow (1, 1) \rightarrow (2, 1) \rightarrow (3, 2) \mapsto (5, 0)$ . Then  $NE(P_5) = 5 + 0 = 5$ .

**Theorem 1** The number of edge cover sets of a path of  $n$  edges, is:  
 $F_n = \text{ClosestInteger} \left[ \frac{1}{\sqrt{5}} \left( \frac{1+\sqrt{5}}{2} \right)^n \right]$ .

Proof. The thread  $(\alpha_i, \beta_i)$ ,  $i = 1, \dots, n$  used for computing  $NE(P_n)$ , coincides with the Fibonacci numbers:  $(F_1, F_0) \rightarrow (F_2, F_1) \rightarrow (F_3, F_2) \rightarrow (F_4, F_3) \dots \rightarrow (F_{n-1}, F_{n-2}) \mapsto (F_n, 0)$ . Then, we infer that  $(\alpha_i, \beta_i) = (F_i, F_{i-1})$  for  $i = 1, \dots, n-1$  and  $\alpha_n = F_n, \beta_n = 0$ . And then  $NE(P_n) = \alpha_n + \beta_n = F_n$ .

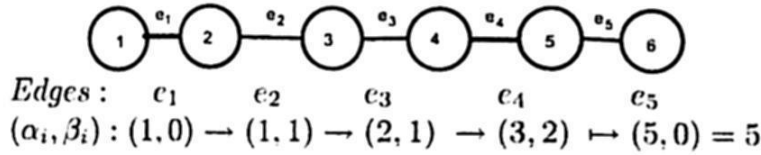


Fig. 2. Counting edge covers on a path

### Case B: Counting Edge Covers on Trees

Let  $T = (V, E)$  be a rooted tree. We distinguish each edge on the tree as follows: *root-edges*, which are the edges with one endpoint in the root node; *leaf-edges*, which are the edges with one endpoint in a leaf node of  $T$ . Given any intermediate node  $v$  of  $T$ , we call a *child-edge* of  $v$  to the edge connecting  $v$  with any of its children nodes, and the edge connecting  $v$  with its father node is called the *father-edge* of  $v$ .  $NE(T)$  is computed traversing by  $T$  in post-order and associating a pair  $(\alpha_e, \beta_e)$  with each edge  $e$  of  $T$ , except for the leaf edges.

**Algorithm #Edge\_Covers\_in\_trees( $T$ )**

**Input:** A rooted tree  $T$  with root vertex  $v_r$

**Output:**  $NE(T)$

**Procedure:**



1. We reduce the input tree  $T$  to other tree  $T'$  by cutting all leaf nodes and leaf-edges from  $T$ , and by labeling as covered nodes all father nodes of the original leaf nodes of  $T$ .
2. Traversing by  $T'$  in post-order, a pair  $(\alpha_e, \beta_e)$  is associated with each edge  $e$  in  $T'$ . Each pair is computed in the following way:
  - (a)  $(\alpha_e, \beta_e) = (1, 1)$  if  $e$  is a leaf-edge of  $T'$ .
  - (b) when an internal node  $v$  is visited and it has a set of child-edges, e.g.  $u_1, u_2, \dots, u_k$  are the child-edges of  $v$ , as we have already visited all child-edges of  $v$  then each pair  $(\alpha_{u_j}, \beta_{u_j})$ ,  $j = 1, \dots, k$  has been computed and associated to the child-edges. The set of child-edges of  $v$  is considered as just one child-edge  $e_u$  and its associated pair  $(\alpha_u, \beta_u)$  is computed, as:

$$\alpha_u = \prod_{j=1}^k (\alpha_{u_j} + \beta_{u_j}) - \prod_{j=1}^k \beta_{u_j}; \quad \beta_u = \prod_{j=1}^k \beta_{u_j} \quad (4)$$

where  $\alpha_u$  carries the number of different combinations of the child-edges of  $v$  for covering  $v$ , while  $\beta_u$  gives the number of combinations among the child-edges of  $v$  which do not cover to  $v$ . The case in which  $v$  has just one child-edge is consider in this case, with  $\alpha_u = \alpha_{u_1}$  and  $\beta_u = \beta_{u_1}$ . The pair associated to the father-edge  $e_v$  of  $v$  is computed as follows:

$$(\alpha_v, \beta_v) = \begin{cases} (\alpha_u + \beta_u, \alpha_u) & \text{if } v \text{ is a free node or,} \\ (\alpha_u + \beta_u, \alpha_u + \beta_u) & \text{if } v \text{ is a cover node} \end{cases}$$

This step (2) is iterated until it computes the pairs  $(\alpha_e, \beta_e)$  for all edge  $e$  of  $T'$  and it stops when it arrives to the root node  $v_r$ .

3. Let  $(\alpha_{u_r}, \beta_{u_r})$  be the pair associated with the root-edge of  $v_r$ .  $NE(T)$  is computed according of the status of  $v_r$ ; if  $v_r$  is a covered node then  $NE(T) = \alpha_{u_r} + \beta_{u_r}$ , otherwise  $NE(T) = \alpha_{u_r}$ .

This procedure returns  $NE(T)$  in time  $O(n+m)$  which is the necessary time for traversing  $T$  in post-order. Notice that the post-order search allows to give an evaluating order for each edge of  $T$  and at the same time, to compute the number of edge covers. We denote with  $\succ$  the application of recurrence (4).

**Example 2** Let  $T$  be an input tree with root vertex  $v_r$  (see fig. 3a).  $T'$  is the reduced tree from  $T$  where its covered nodes are marked by a black point inside of the nodes (see fig. 3b).  $T'$  is traversing in post-order and each pair  $(\alpha_e, \beta_e)$  is associated with each edge  $e$  of the tree. The pairs for the child-edges of  $v_r$ , are:  $(1,1)$ ,  $(4,3)$  and  $(6,3)$ . Those three edges are combined in only one edge  $e_r$  by applying recurrence (4), the resulting pair  $(\alpha_r, \beta_r)$  was computed as:  $\alpha_r = (1+1) * (4+3) * (6+3) - 1 * 3 * 3 = 117$  and  $\beta_r = 1 * 3 * 3 = 9$ . Since  $v_r$  is the root node and it is free, then  $NE(T) = \alpha_r = 117$ .

### Case C: Counting Edge Covers on Simple Cycles

Let  $C_n = (V, E)$  be a simple cycle with  $n$  edges. Let us order the nodes and edges of  $C_n$ , as:  $V = \{v_1, \dots, v_n\}$  and  $E = \{e_1, \dots, e_n\}$ ,  $e_i = \{v_i, v_{i+1}\}$ ,  $i = 1, \dots, n-1$ ,  $e_n = \{v_n, v_1\}$ .

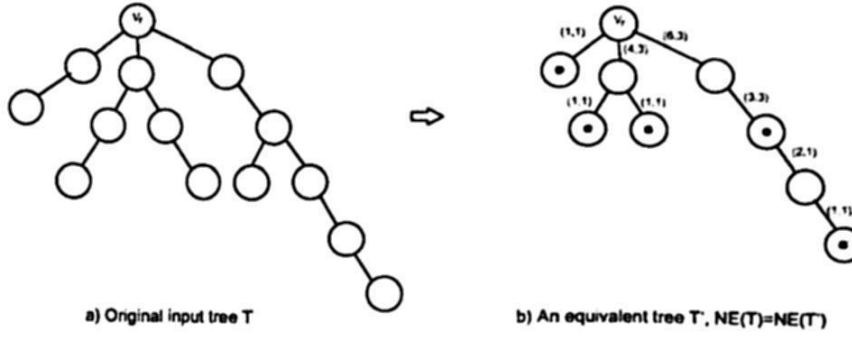


Fig. 3. Computing the number of edge covers for a tree

A computing thread  $L_p: (\alpha_i, \beta_i), i = 1, \dots, n$  is used for counting the edge covers of the path  $P_n$  contained in  $C_n$ . A depth-first search starts in the edge  $e_1$  and the pair  $(\alpha_1, \beta_1) = (1, 1)$  is associated with it, since  $e_1$  is not a fixed edge.

Since all nodes in  $C_n$  have degree two, when a new edge is visited during the depth-first search the Fibonacci recurrence (1) is applied. Then, after  $n$  applications of recurrence (1), the pair  $(\alpha_n, \beta_n) = (F_{n+1}, F_n)$ ,  $F_i$  being the  $i$ -th Fibonacci number, is obtained. Let  $\mathcal{CE}(NC_n) = \{\mathcal{E} \subseteq E : \mathcal{E} \text{ is one of the edge sets counted through the thread } L_p\}$ , then  $NC_n = |\mathcal{CE}(NC_n)| = \alpha_n + \beta_n = F_{n+2}$ .

There are some edge sets  $\mathcal{E} \in \mathcal{CE}(NC_n)$  where the edges  $e_1$  and  $e_n$  do not appear, since the computation of the thread  $L_p$  allow these cases, although when the values  $\beta_1 = 1$  and  $\beta_n > 0$  represent not edge covers for  $C_n$  because they do not cover the node  $v_1$ . Then, in order to count only the edge covers of  $C_n$ , we have to subtract from  $NC_n$  the edge sets that do not cover  $v_1$ .

Let  $Y = \{\mathcal{E} \in \mathcal{CE}(NC_n) : e_1 \notin \mathcal{E} \wedge e_n \notin \mathcal{E}\}$  be the edge sets which cover all nodes of  $C_n$  except  $v_1$ , then  $NE(C_n) = NC_n - |Y|$ . In order to compute  $|Y|$  a new computing thread  $(\alpha'_i, \beta'_i), i = 1, \dots, n$ , denoted by  $C'_n$ , is built.  $C'_n$  starts with the pair  $(\alpha'_1, \beta'_1) = (0, 1)$  since in this way we consider the edge sets where  $e_1$  does not appear. After  $n$  applications of recurrence (1), we obtain as last pair of  $C'_n$ ,  $(\alpha'_n, \beta'_n) = (F_{n-1}, F_{n-2})$ . For considering only the edge sets where neither  $e_1$  nor  $e_n$  appear,  $(\alpha'_n, \beta'_n)$  is taken as  $(0, \beta'_n) = (0, F_{n-2})$ , and  $|Y| = F_{n-2}$ . Then,  $NE(C_n) = NC_n - |Y| = F_{n+2} - F_{n-2}$  and we deduce the following theorem.

**Theorem 2** *The number of edge cover sets of a simple cycle  $C_n$  with  $n$  edges, expressed in terms of Fibonacci numbers, is:  $NE(C_n) = F_{n+2} - F_{n-2}$ .*

In the following examples, we denote with ' $\curvearrowright$ ' the binary operation between pairs:  $(\alpha_n, \beta_n)$  and  $(\alpha'_n, \beta'_n)$  - the final pairs of the two computing threads of a cycle - whose result is the pair:  $(\alpha_n, \beta_n - \beta'_n)$  and which has to be associated with the last edge  $e_n$  of a cycle  $C_n$ . Notice that the computation of  $NE(C_n)$  has a time complexity of  $O(n)$  since we compute the two threads:  $L_p$  and  $C'_n$  in parallel while the depth-first search is applied.

**Example 3** *Let  $C_6$  be the simple cycle illustrated in figure 4. Applying theorem (2), we have that  $NE(C_6) = F_{6+2} - F_{6-2} = F_8 - F_4 = 21 - 3 = 18$ .*

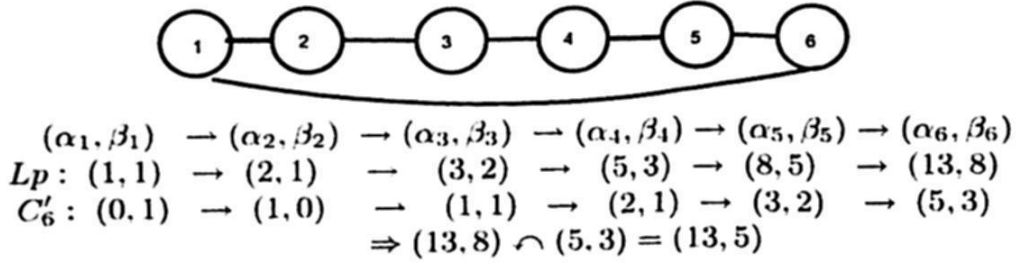


Fig. 4. Obtaining the number of edges covers of a cycle

#### 4 Counting Edge Cover Sets for General Graphs

Let  $G = (V, E)$  be a connected graph with  $|V| = n$ ,  $|E| = m$  and such that  $\Delta(G) \geq 2$ . Choose a node  $v_r \in V$  (e.g. the node with minimum degree in  $V$ ) for starting a depth-first search over  $G$  in order to build a spanning tree  $T_G$  where  $v_r$  is the root node.

The edges in  $T_G$  are called *tree edges*. While the edges in  $(E - E(T_G))$  are called *back edges*. Given a back edge  $e$ , the union of the path in  $T_G$  between the endpoints of  $e$  and the same edge  $e$  forms a simple cycle, such cycle is called a *fundamental cycle* of  $G$  with respect to  $T_G$ . Then, each back edge embraces the maximum path contained in a fundamental cycle. Let  $\mathcal{C} = \{C_1, \dots, C_k\}$  be the set of fundamental cycles found during the depth first search of  $G$ . Given any pair of fundamental cycles  $C_i$  and  $C_j$  from  $\mathcal{C}$ , if  $C_i$  and  $C_j$  share edges, we call them *intersecting cycles*; otherwise, they are called *independent cycles*.

We call *critical point* of the graph  $G$  to each incident node  $v_p$  to a back edge. Given a critical point  $v_p$  of  $G$  the set of its incident edges  $E_p \subset E$  is called a *critical edge set*. Before presenting our most general counting algorithm, let us show the method used for processing critical edge sets of a graph.

##### Case D: Processing Critical edge sets of a Graph

A main computing thread  $L_p : (\alpha_i, \beta_i)$ ,  $i = 1, \dots, m$ , is used for counting the edge cover sets of an input graph  $G$ . When an edge  $e_i \in G$  is visited for the first time during the depth-first search (called the current edge), its associated pair  $(\alpha_i, \beta_i)$  is computed according to the corresponding recurrence with the type of edge that  $e_i$  is in  $G$ . However, the series  $L_p$  counts all subsets of a critical edge set  $S$  of  $G$  including the case when all edges of  $S$  do not appear. Although for this latter case (for the empty subset of  $S$ ) the critical point associated with  $S$  is not covered.

Then, when a critical point  $v_p$  is visited for first time, we have to open a new computing thread in order to count the number of subsets considered by  $L_p$  and where all edges of  $E_p$  do not appear, and such number has to be subtracted from the current value of  $L_p$ . Notice that this case is a generalization of the processing of a back edge for simple cycles. The computing thread  $L_p$  is always active until



the counting process finishes, while the subordinated threads are active until all edges of their corresponding critical edge set have been visited.

If a new critical point  $u_p$  is visited before visiting all edges of a previous critical edge set  $E_p$ , then some new auxiliary threads are created in order to count the number of subsets counted by the current computing threads and for which all edges of the new critical edge set  $E_u$  do not appear. In fact, for each active thread  $l_i$  (called a main thread of  $E_u$ ) one auxiliary thread subordinated to  $l_i$  is created with initial pair  $(0, \beta_u^{l_i})$  being  $\beta_u^{l_i}$  equals to  $\beta_u$  in its respective main thread  $l_i$ .

When all edges of a critical edge set  $E_p$  had already visited, then its associated critical point  $v_p$  has been processed and the binary operator  $\curlywedge$  is applied between the main and its corresponding subordinated thread. After applying  $\curlywedge$ , the subordinated threads created for processing  $E_p$  are closed and they stop to be active, keeping only the main threads of  $E_p$ .

We illustrate in figure 5, how to process critical edge sets. The initial graph  $G$  is formed by a cycle union a path on each endpoint of the back edge. The method used for processing the two critical edge sets of  $G$  is representative of the way of processing critical edge sets. For this graph,  $e_8$  is the unique back edge. The two critical points in  $G$  are  $v_4$  and  $v_8$  which are the endpoints of the back edge. The two critical edge sets are:  $E_1 = \{e_3, e_4, e_8\}$  and  $E_2 = \{e_7, e_8, e_9\}$ .

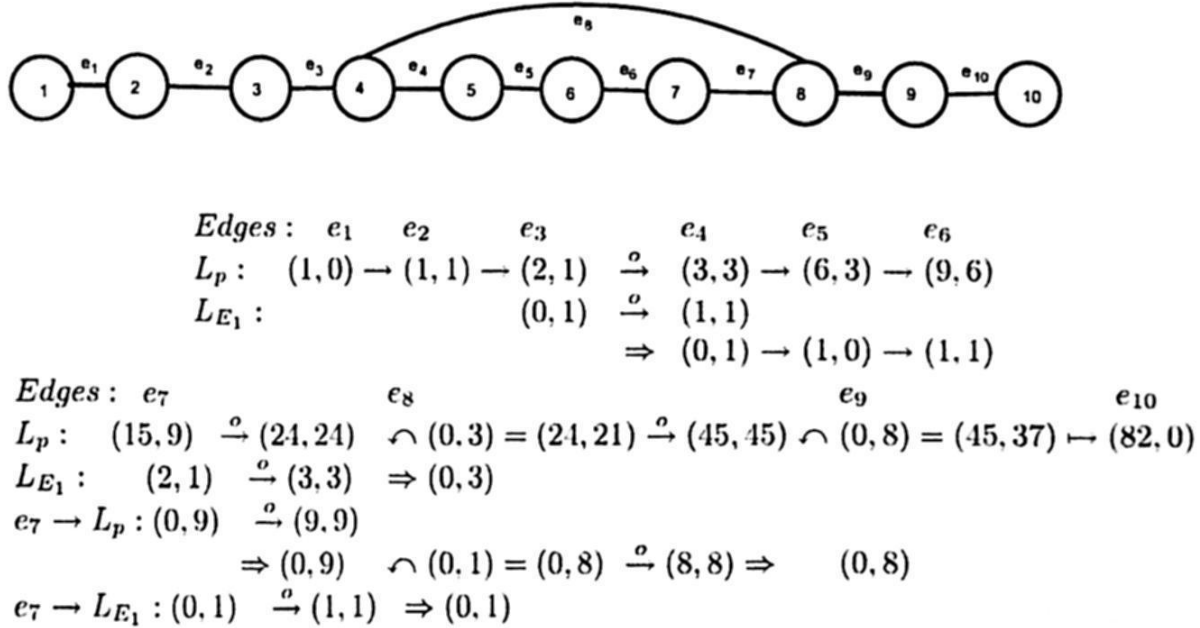


Fig. 5. Computing  $NE(G)$  on a cycle combined with a path

The main thread  $L_p$  starts with  $(\alpha_1, \beta_1) = (1, 0)$  since  $e_1$  is a fixed edge. The recurrence (1) is applied for computing  $(\alpha_2, \beta_2)$  and  $(\alpha_3, \beta_3)$  since  $v_2$  and  $v_3$  have degree 2. Since  $e_3$  is member of the critical edge set  $E_1$ , we have to count the edge sets not containing  $E_1$  and which are being considered by  $L_p$ . Then, the

auxiliary thread  $L_{E_1}$  is created and it is subordinated to  $L_p$ . The initial pair for  $L_{E_1}$  is  $(\alpha'_3, \beta'_3) = (0, \beta_3) = (0, 1)$  since  $\beta_3 = 1$  in  $L_p$ .

When  $e_4$  is visited, recurrence (3) is applied over each active threads:  $L_p$  and  $L_{E_1}$ , since  $\delta(v_4) > 2$ . Although, as we want to count in  $L_{E_1}$  only the edge sets where  $e_3, e_4$  (and in advance  $e_8$ ) do not appear, then  $(\alpha'_4, \beta'_4)$  is changed to  $(0, \beta'_4)$ , and in this case  $(\alpha'_4, \beta'_4) = (0, 1)$ .

Next, recurrence (1) is applied until arriving to edge  $e_7$  since the nodes  $v_5, v_6$  and  $v_7$  have degree two. As  $e_7$  is member of the critical edge set  $E_2$ , we must count the edge sets not containing  $E_2$  and which are considered by the active threads. Then, two new auxiliary threads are created. The thread denoted by  $e_7 \rightarrow L_p$  which is subordinated to  $L_p$  and  $e_7 \rightarrow L_{E_1}$  which is subordinated to  $L_{E_1}$ . Their corresponding starting pairs are:  $(0, 9)$  and  $(0, 1)$  since 9 and 1 are the edge sets counted by  $L_p$  and  $L_{E_1}$  respectively, where  $e_7$  does not appear.

When  $e_8$  is visited, the recurrence (3) is applied over each active computing thread since  $\delta(v_8) > 2$ . Since all edges of  $E_1$  have been visited, meaning that the first critical point of  $G$  has been processed, the operator  $\curlywedge$  is applied between  $L_p$  and  $L_{E_1}$ , as well as between  $e_7 \rightarrow L_p$  with  $e_7 \rightarrow L_{E_1}$ . After to apply the operator  $\curlywedge$ , the threads  $L_{E_1}$  and  $e_7 \rightarrow L_{E_1}$  are closed and they stop to be active.

As the thread  $e_7 \rightarrow L_p$  has been used for counting the sets not containing the edges of  $E_2$ , then its corresponding pair  $(\alpha'_8, \beta'_8)$  is changed to  $(0, \beta'_8)$ . When  $e_9$  is visited, the recurrence (3) as well as the operator  $\curlywedge$  are applied between the two active threads since all edges of  $E_2$  have already been visited, remaining only the main thread  $L_p$ . Finally, when  $e_{10}$  is visited, the recurrence (2) is applied and the last pair is  $(\alpha_{10}, \beta_{10}) = (82, 0)$ . And then,  $NE(G) = \alpha_{10} + \beta_{10} = 82$ .

One of the main techniques for counting objects on a graph has been the Markov chain Monte Carlo method [1, 2, 5, 3]. Although it is likely that this approximation technique provides efficient algorithms only for graphs with bounded maximum degree. For example, the Markov chain Monte Carlo procedures for counting the number of independent sets of a graph is likely to fail for graphs of maximum degree six or higher [5]. In our case, we are presenting deterministic and exact procedures for counting the number of edge covers of a graph based on the topological structure of the graph and without consideration of its maximum degree. Note that the basic procedures (A), (B) and (C) allow us to compute the number of edge cover sets for any graph with maximum degree 2.

## 5 Estimating the Relevance of Communication Lines

Complex networks, modeled as large graphs, received much attention during these last years. However, topological information on these networks is only available through intricate measurement procedures. Until recently, most studies assumed that these procedures eventually lead to samples large enough to be representative of the whole, at least concerning some key properties. An important application of counting edge covers is for estimating the degree of reliability in communication networks [10].

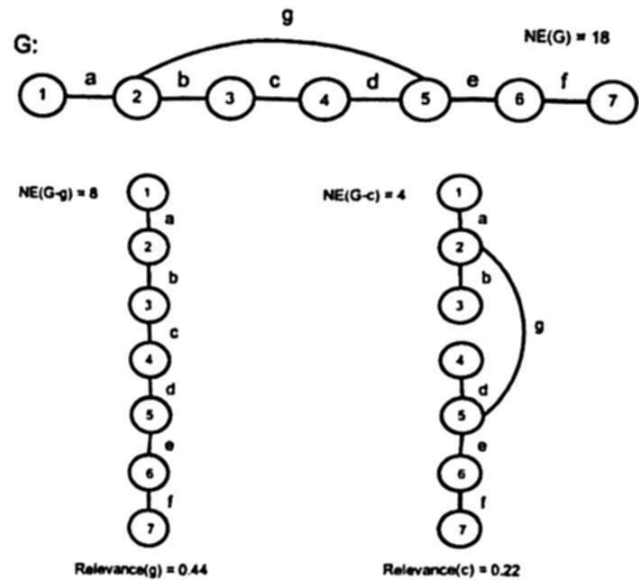


Fig. 6. Estimating the relevance of lines g and c

Table 1. Estimating the relevance of the lines of G

Without line	Edge covers	Relevance
a	15	0.17
b	7	0.61
c	4	0.77
d	6	0.66
e	8	0.55
f	10	0.44
g	8	0.55

For example, if we assume that the communication lines (edges) in a network  $G$  has the same 'failure probability' and those failures are independent of one another, we can measure different classes of reliability of the network, given that an edge  $c \in G$  fails, according to what component of the network is considered. A way to estimate the 'relevance' of a line  $c$  in the network  $G$  is by applying the conditional probability  $P_{c/G}$  which can be approximated by the fraction of the number of edge covers which are subtracted when the edge  $c$  is removed (fails), that is,  $P_{c/G} = 1 - \frac{NE(G-c)}{NE(G)}$ . Thus,  $P_{c/G}$  gives the strategic value of an edge  $c$  in a network  $G$  by estimating the relevance of the line. As  $c$  is any edge of  $G$  then  $P_{c/G}$  could be used for estimating the relevance for any edge of  $G$ .

Then, the measure  $P_{c/G}$  could be used for estimating the strategic value of any edge  $c$  of a graph  $G$  with respect to the other lines in  $G$ . In such a way that for greater values of  $P_{c/G}$  means that the line  $c$  is most relevant for maintaining the connectivity of  $G$ , in case of failures, with respect to the other lines in  $G$ .

## 6 Conclusions

We determine different recurrence relations for counting, in incremental way, the number of edge cover sets of a graph according with each node and edge of the graph is being visited for the first time during a depth-first search. If the input graph has simple topologies, like: paths, trees, simple cycles or combination of the previous topologies, we can compute the number of edge covers in linear time in the size of the graph.

The class of graphs for which our novel efficient algorithms compute their number of edge cover sets determines a class of polynomial instances for counting the number of edge covers, and such class is a superclass of graphs of degree two without restrictions on the degree of the graphs, but rather, it depends on the topological structure of the graphs.

To know how to compute the number of edge covers is helpful for estimating the reliability of a communication network. For example, we have shown how to estimate the 'relevance' of any line  $c$  of the network based on the proportion of the number of edge covers where  $c$  does not appear with respect to the total number of edge covers in the network.

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