Succinct Representations of Planar Maps

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Abstract

This paper addresses the problem of representing the connectivity information of geometric objects using as little memory as possible. As opposed to raw compression issues, the focus is here on designing data structures that preserve the possibility of answering incidence queries in constant time. We propose in particular the first optimal representations for 3-connected planar graphs and triangulations, which are the most standard classes of graphs underlying meshes with spherical topology. Optimal means that these representations asymptotically match the respective entropy of the two classes, namely 2 bits per edge for 3-connected planar graphs, and 1.62 bits per triangle or equivalently 3.24 bits per vertex for triangulations. These representations support adjacency queries between vertices and faces in constant time.

Keywords: Succinct data structures, graph encoding, data compression, planar maps, geometric data structures, triangulations.

1 Introduction

1.1 Connectivity vs geometry

A geometric object is often represented by a polygonal mesh which contains two kinds of information: the geometry and the connectivity. The connectivity is a graph which describes how vertices are linked by edges and faces, while the geometry consists in the vertices coordinates. In usual representations such as VRML format or pointers representations in main memory [7], the connectivity is the most expensive part: it costs hundreds of bits per vertex while the geometry costs only tens of bits per vertex. As a matter of fact, in such formats where the connectivity is represented by numbering the vertices and giving indexes, the cost of connectivity has order $\Theta(n \log n)$. As done in previous existing works

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on compact representations [9, 18, 6], we shall concentrate in this paper on reducing the connectivity cost, and leave the problem of reducing the geometry cost aside. We observe however that our structure is compatible with several of the standard approaches to geometry compression (point coordinates can be given in a local framework).

1.2 Compression vs succinct structures

A geometric object can be represented either in a linear format for disk storage or network transmission, or stored in main memory for the exploration of the object. In the first case, reducing the size is called compression, and compressing the connectivity of various kind of meshes has been successfully attacked in recent years [20, 21, 14, 23, 22, 10, 15] (for a survey of the most recent advances in this field we refer to [5]). In this paper we deal with the second case of main memory representation, and our aim is to design a data structure having a small size and allowing to answer queries in constant time. Usual queries consist in going from a face to its neighbor or asking if two vertices are adjacent in the mesh. Such a structure is called succinct if the cost asymptotically matches the entropy of the class and compact if it matches it up to a constant factor. More precisely, given a class of objects with a size parameter $n$ (e.g. the number of elements of some kind), we consider the number of objects of size $n$ in the class. If this number has an exponential growth of order $2^{\alpha n}$ when $n$ goes to infinity, the entropy\footnote{As done in related works on compact encodings, we use the term entropy to indicate the information theoretic lower bound. Let us observe that we will consider classes of graphs with uniform distribution.} of the class is defined to be $\alpha n$. A representation is then compact if it uses $O(n)$ bits and succinct if it uses $\alpha n + o(n)$ bits. Observe that a correct representation cannot use less than $\alpha n$ bits for it must be possible to distinguish all objects.

1.3 General framework

The general framework we use for designing a compact or succinct data structure for a given class of objects of size $n$ is sketched here:

- First the object is split into tiny pieces of size $O(\log n)$, tiny meaning small enough so that a catalog of all possible pieces can be constructed in $o(n)$ time and space. Then a tiny piece is represented by its index in the catalog, and the sum of the sizes of all indexes is expected to match the entropy of the class.

- The incidence relations describing how the splitting into tiny pieces has been done is encoded in a graph $G$ of tiny pieces. Since there are $O\left(\frac{n}{\log n}\right)$ tiny pieces with a linear number of incidences between them, a classical representation of this graph using pointers of logarithmic size costs $O(n)$ and this approach already yields a compact data structure.
• Small pieces of $O(\lg^2 n)$ size are constructed by joining $\lg n$ tiny pieces, this allows to use pointers of size $O(\lg n)$ only between small pieces while adjacencies between tiny pieces are described with local pointers of size $O(\lg \lg n)$. Since the number of small and tiny pieces are respectively $O(\frac{n}{\lg n})$ and $O(\frac{\lg n}{\lg n})$, this multi-level approach yields sublinear costs of $O(\frac{n}{\lg n} \lg 2 n)$ and $O(\frac{\lg n}{\lg n} \lg n)$ for $G$, making the structure succinct.

1.4 Related results

We briefly review in this section a few results about representations of graph connectivity for 3-connected planar graphs and triangulations. These results can be given in terms of $n$ the number of vertices, $m$ the number of faces or $e$ the number of edges. In the special case of triangulations of a topological sphere, $6n \simeq 2e = 3m$ and the entropy is $1.62m = 3.24n$ bits [24]. For planar triangulations with a boundary, the entropy is $2.175m$ bits. For general 3-connected planar graphs the entropy is $2e$ bits [25].

On the practical side, classical main memory representations use pointers: $n + 6m$ pointers are needed for triangulations and $n + 6e$ for 3-connected graphs [7], where a pointer means 32 bits with real pointers and $\lg n$ bits using indexes. A cheaper solution with $2e$ pointers [16] has been proposed with the price of a higher access cost to neighbors. None of these $O(n \lg n)$ structures are compact.

The above framework has been introduced for balanced parenthesis words (Dyck words) by Jacobson [13] for the compact representation, and by Munro and Raman [18] for the succinct representation.

The size parameter of a parenthesis word is its number of characters and optimality means 1 bit per character. In this context the natural query is to ask for the matching parenthesis of the parenthesis at a given position. These results on parenthesis words allow for succinct representations of plane (aka ordered) trees using $2n$ bits. Then a planar map can be decomposed into several trees which can be succinctly represented.

However, this transformation from graphs to trees being non bijective, it yields representations for planar graphs that are not succinct but only compact. Along these lines, a representation of planar graphs using $2e + 8n$ bits was given [18] and then improved to $2e + 2n$ bits [9, 8].

In our previous work [1], we have shown how to extend the framework so that it can be applied directly to triangulations with a boundary and provided a succinct representation for this (larger) class of triangulations. This approach was also successful in dealing with local dynamic updates of triangulations of higher genus surfaces [2] (a comparison with previous works is shown in Table 1).

With a slightly different strategy, Blandford et al. [6] showed how to design a compact data structure supporting adjacency and degrees queries on vertices for special classes of graphs having small separators. However this approach needs efficient algorithms for finding separators and the exact cost of the representation is difficult to characterize. As in previous works, our model of computation is a RAM machine, with $O(1)$ time access on words of size $\Theta(\lg n)$. 

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Table 1: Comparison of existing compact representations for simple planar graphs, with $e$ edges, $m$ faces and $n$ vertices (lower order terms are omitted).

<table>
<thead>
<tr>
<th>Algorithm</th>
<th>triangulated</th>
<th>3-connec.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Munro Raman (FOCS 97)</td>
<td>$2e + 8n$ or $7m$</td>
<td>$2e + 8n$</td>
</tr>
<tr>
<td>Chuang et al. (ICALP 98)</td>
<td>$2e + n$ or $3.5m$</td>
<td>$2e + 2n$</td>
</tr>
<tr>
<td>Chiang et al. (SODA 01)</td>
<td>$2e + 2n$ or $4m$</td>
<td>$2e + 2n$</td>
</tr>
<tr>
<td>Castelli Aleardi et al. (WADS 05)</td>
<td>$2.175m$</td>
<td>-</td>
</tr>
<tr>
<td>our new encodings</td>
<td>$1.62m$</td>
<td>$2e$</td>
</tr>
</tbody>
</table>

1.5 Contribution

The contribution of this paper is twofold. As far as results are concerned, we propose the first succinct data structures for representing planar triangulations without boundary (triangulations of a topological sphere) and 3-connected planar maps, as stated in Theorem 1 below.

**Theorem 1.** There exist succinct representations

- of planar triangulations (without boundaries) requiring asymptotically $1.62$ bits per triangle,
- of 3-connected planar graphs requiring asymptotically $2$ bits per edge.

Both representations support local standard navigation in $O(1)$ time, using an extra storage of size $O(\frac{n \log \log n}{\log n})$.

From the methodological point of view, we formalize the catalog-tiny-small framework. With respect to the seminal data structures proposed for words and trees [13, 18], and for triangulations in our paper [1], the present framework makes explicit the local planarity properties on which the approach is based.

Furthermore, it relaxes the property, central in those previous works, that tiny pieces should be taken in a class with the same entropy as the main class of objects represented. Finally, in order to develop our two new applications (triangulations and 3-connected planar maps), we design new splitting schemes.

Next section will formalize the catalog-tiny-small framework, while Section 3 describes its use for planar maps and Section 4 for triangulations.

2 The catalog-tiny-small framework

In this section we make gradually more precise the general framework sketched in Section 1. At a first level of details, the framework applies to any data structure which has linear entropy and can be decomposed into regions connected by a globally sparse graph: this includes parenthesis words, trees, and graphs on surfaces. The framework is then specialized to connectivity structures of meshes.
2.1 Additivity of the entropy and the catalog

In order to apply our framework to a class of combinatorial objects, we first need that the entropy be linear. More precisely consider a class $C = (C_n)$ of objects where $n$ is intended as a size parameter (the number of some elementary cells) and assume that the set $C_n$ of objects of size $n$ has a finite cardinality $|C_n|$. The entropy $||C_n||$, defined by

$$||C_n|| := \log |C_n|$$

with $\log(x) = \lceil \log_2(x+1) \rceil$, measures the diversity of the class. A class has linear entropy if there exists a constant $\alpha$ such that

$$||C_n|| = \alpha n + o(n) \quad \text{when } n \text{ goes to infinity.}$$

In other terms, the cardinality of the class $C_n$ grows roughly like a simple exponential $2^{\alpha n}$, and $\alpha n$ bits are needed to index an arbitrary element. The constant $\alpha$ is sometimes called the entropy per size unit: $\alpha = 2$ bits per node for binary trees, and, as indicated above, $\alpha = 1.62$ bit per triangles for triangulations, and $\alpha = 2$ bits per edge for 3-connected planar graphs. Observe however that some classes, like the classes of permutations of $\{1, \ldots, n\}$ or of general $n$-vertex graphs have non linear entropies: of order $\Theta(n \log n)$ and $\Theta(n^2)$ respectively.

We intend to decompose each object of $C$ into pieces taken from a catalog of smaller objects: as opposed to previous works, we do not require that this catalog contains elements of $C$, but rather that it contains elements of a class $D = (D_{m,k})$, such that there exists a constant $\beta$ and a positive function $g(m) = O(\log m)$ such that

$$||D_{m,k}|| \leq \alpha m + \beta k \log m + g(m) \quad (1)$$

and $|D_{m,k}| = 0$ if $k \geq Km$ (for some constant $K$).

The objects of $D_{m,k}$ are intended to be used to describe tiny pieces of elements of $C$, with $m$ the number of elementary cells, and $k$ a parameter, called the number of sides, describing the complexity of the boundary of the piece. In the above bound the constant $\alpha$ is expected to be the same as for $C_n$, and $\alpha m$ is the additive part of the entropy, while $\beta k \log m$ is the extra amount of entropy due to the splitting into tiny pieces. By definition of the entropy, $||D_{m,k}||$ bits are sufficient to index an element of $D_{m,k}$ in a table representing all those elements.

We assume more precisely that each element $M$ of $C_n$ can be decomposed into

- a collection $(M_1, \ldots, M_p)$ of elements of $D$, with $M_j \in D_{n_j, k_j}$ for some $n_j, k_j$,

- and an oriented graph $G$ with vertex set $\{N_1, \ldots, N_p\}$, describing how the tiny pieces $\{M_1, \ldots, M_p\}$ are glued together to form $M$, in terms of adjacencies between sides of the $M_j$. 

More precisely, a vertex $N_j$ of $G$ contains the following information:

- $n_j$, $k_j$, and the index of $M_j$ in $D_{n_j,k_j}$,
- indexes to the neighbors of $N_j$ in $G$ (for each side of $M_j$, the index of the corresponding neighbor and side).

In particular the number of edges in the graph $G$ is bounded by the total number of sides in the $M_j$.

We first need an hypothesis ensuring that the additive part of the entropy matches the entropy $\alpha n$ of the class to which $M$ belongs.

**Hypothesis 1.** The decomposition is additive in the size parameter (elementary cells are not shared):

\[ n_1 + \ldots + n_p = n + O\left(\frac{n}{\lg n}\right). \]

As expressed by the term $O\left(\frac{n}{\lg n}\right)$ we allow a negligible number of elementary cells to be shared between tiny pieces.

Observe that we do not require the class $D_m = \bigcup_k D_{m,k}$ to have the same entropy as $C_m$: in particular in our two main examples below we will have $||D_m|| \sim \alpha' m$ with $\alpha' > \alpha$. As a consequence, in order for the representation to be compact we need a second hypothesis on the number of sides.

**Hypothesis 2.** The decomposition involves a sub-linear number of sides:

\[ k_1 + \ldots + k_p = O\left(\frac{n}{\lg n}\right). \]

This second hypothesis implies that the whole cost of storing indexes to all the $M_i$ remains of order $\alpha n$. Next hypothesis ensures that the elements of $D$, needed in the decomposition, fit in a small catalog.

**Hypothesis 3.** In the decomposition each $M_j$ can be taken of size between $\frac{c}{3} \lg n$ and $c \lg n$, where $c < 1/\alpha'$, with $\alpha'$ the entropy per size unit of $D$.

Indeed assume that the indexes are pointing into a table $A$, containing the explicit representations of all elements of $D_m$ for $m \leq c \lg n$ for some constant $c$ (where $c$ depends on the cardinality of $D$). If the constant $c$ is chosen small enough, the number of entries in the table is sub-linear: indeed $||D_m|| = \alpha' m \leq c\alpha' \lg n$, so with $c < 1/\alpha'$, the number of entries in the table is $O(n^{c\alpha'})$. The total storage cost of Table $A$ then remains sub-linear as long as the information for each piece is polynomial in its size $m = O(\log n)$. In particular, explicit answers to local queries (as local adjacency or degree queries on elementary cells) can be stored using auxiliary information without affecting significantly the overall size of the representation.

**2.2 Compactness**

Hypothesis 2 above ensures that the graph $G$ has $O(n/\lg n)$ edges, and Hypothesis 3 that it has $O(n/\lg n)$ vertices. This is already enough to obtain compactness.
Lemma 2. Under Hypotheses 1, 2, and 3, the storage of the graph \( G \) requires \( O(n) \) bits.

Proof. Recall that each vertex \( N_j \) of \( G \) contains: \( n_j, k_j \) and the index of \( M_j \) in \( D_{n_j,k_j} \), as well as indexes to the neighbors of \( N_j \) in \( G \). As \( n_j \) and \( k_j \) are smaller than \( c \lg n \) and \( Kc \lg n \) respectively, we can store them in \( 2\lg \lg n + O(1) \) bits. When summing over the \( O(n/\lg n) \) vertices of \( G \) we get a \( O(\frac{n \lg \lg n}{\lg n}) \) bits cost. Using Equation (1) the index of \( M_j \) requires \( \alpha n_j + \beta k_j \lg n_j + O(\lg n_j) \) bits. Summing over all \( M_j \) yields a global cost of

\[
\sum_j \|D_{n_j,k_j}\| \leq \alpha \sum_j n_j + \beta \sum_j (k_j) \lg(\max_j n_j) + O\left( \sum_j \lg n_j \right) \\
\leq \alpha n + O\left( \frac{\alpha n}{\lg n} \right) + \beta O\left( \frac{n}{\lg n} \right) \lg(\lg n) + \frac{n}{\lg n} O(\lg \lg n)
\]

Using Hypotheses 1 and 2, the cost of all indices to Table A thus reduces to \( \alpha n + O\left( \frac{n \lg \lg n}{\lg n} \right) \) bits. Since there are \( O(n/\lg n) \) vertices in \( G \), the index of a neighbor uses \( \lg n + O(1) \) bits. Vertex \( N_j \) has \( O(k_j) \) neighbors, thus the total cost for storing indexes to the neighbors is \( (\lg n + O(1)) \cdot \sum_j k_j = O(n) \). Summing all these components yields the claimed complexity. \( \square \)

2.3 Succinctness

In the proof of Lemma 2 the linear part of the storage came from two kinds of contributions: the contribution of indexes in the catalog which is dominated by the entropy \( \alpha n \), and the contribution of the neighboring relations in the graph \( G \). In this section, the cost of this second part is reduced to be sub-linear.

The graph \( G \) is partitioned into small pieces gathering \( O(\lg n) \) tiny pieces. More precisely we assume that we are able to construct a graph \( G' \) obtained by merging several vertices of \( G \) in a vertex of \( G' \) and linking two such vertices if there is an edge in \( G \) between two of their elements. The vertex set of \( G' \) is \( \{N'_1, N'_2, \ldots, N'_{p'}\} \) and we denote \( |N'_i| \) the number of vertices of \( G \) that have been merged to obtain \( N'_i \) and \( \deg'(N'_i) \) the degree of \( N'_i \) in \( G' \).

A vertex \( N'_i \) of \( G' \) then consists of the following information:
- \( |N'_i| \) and \( \deg'(N'_i) \)
- the information for all the vertices \( N_j \in N'_i \), stored in a single memory zone.
- indexes to neighbors of \( N'_i \) in \( G' \).

The graph \( G' \) has moreover to satisfy the following hypotheses.

Hypothesis 4. The number of tiny pieces gathered in each small piece \( N'_i \) satisfies: \( \frac{1}{2} \lg n \leq |N'_i| \leq \lg n \).

The number of edges from a given vertex of \( G' \) can be bounded by summing over its elements:

\[
\deg'(N'_i) \leq \sum_{N_j \in N'_i} \deg(N_j) = \sum_{N_j \in N'_i} k_j \leq |N'_i|Kc \lg n = O(\lg^2 n)
\]
(where $K$ is a constant describing the size of the boundary of tiny pieces, as introduced at Equation (1)). But we need a stronger hypothesis on the total number of edges of $G'$.

**Hypothesis 5.** The number of edges of $G'$ is linear in its number of vertices:

$$\deg'(N_1') + \ldots + \deg'(N_p') = O\left(\frac{n}{\lg^2 n}\right).$$

Observe that Hypothesis 4 and the way we gather the nodes of graph $G$ guarantee that the number of nodes of $G'$ is at most $O\left(\frac{n}{\lg n}\right)$. Next Lemma ensures that the storage cost for graph $G$ is asymptotically negligible.

**Lemma 3.** Under Hypotheses 1, 2, 3, 4 and 5 the graph $G$ can be stored using $\alpha n + O\left(\frac{n\lg n}{\lg^2 n}\right)$ bits.

**Proof.** First we notice that in the above representation of $G$ (Lemma 2), a vertex $N_j$ uses $O(\lg^2 n)$ bits ($O(\lg n)$ bits for the index of $M_j$ in the relevant catalog and $O(\lg n)$ for each of its, at most, $Kc\lg n$ neighbors) which gives easily a bound of $O(\lg^3 n)$ for the memory needed by all vertices of $G$ that merge in some $N_i'$. Hence a local reference to memory addresses relative to some known $N_i'$ requires $\lg \lg n + O(1)$ bits. Let us now return to the information stored in a vertex $N_j \in N_i'$ of $G$. To refer to a neighbor $N_l$ of $N_j$, instead of using an address in the whole memory devoted to $G$, we refer first to the vertex $N_k' \ni N_l$ and then give the address of $N_l$ in the memory zone devoted to elements of $N_k'$. Referring to $N_k'$ is done indirectly by giving its index in the array of the, at most, $O(\lg n)$ neighbors of $N_i'$. Thus a reference to a neighbor $N_l$ of $N_j$ costs $O(\lg \lg n)$ bits. The analysis of the size of $G$ is similar to the argument used in Lemma 2: here the cost of a reference to a neighbor does go from $O(\lg n)$ down to $O(\lg \lg n)$ which yields the claimed complexity. The additional cost for $G'$ is sublinear since it has $O\left(\frac{n}{\lg^2 n}\right)$ vertices by Hypothesis 4 and edges by Hypothesis 5 and each costs $O(\lg n)$ bits.

### 2.4 Local planarity and mesh connectivity

The above framework easily applies to trees: a tree with $n$ vertices can be recursively decomposed into tiny trees of logarithmic size, with sides consisting of edges connecting nodes of different tiny trees.

More interestingly Hypotheses 1, 2, 3, 4 and 5 are naturally satisfied when dealing with mesh connectivity (that is, for maps on surfaces). Following for instance [1], a triangulation $M$ with $n$ faces can be partitioned into tiny regions by decomposing an arbitrary dual spanning tree into tiny trees (that is, a spanning tree of the dual graph, connecting all faces across edges): upon reforming a (local) planar triangulation from each tiny tree, a decomposition $(M_1, \ldots, M_p)$ is obtained, such that:

- each tiny triangulation contains between $\frac{1}{12} \lg n$ and $\frac{1}{4} \lg n$ triangles (that is, $c = 1/4$);
- each triangle belongs to exactly one tiny triangulations;
- boundary edges are regrouped into sides (sequences of edges separating it from the same tiny triangulation).

The bound on the number of edges of the graph $G$ that describes adjacency relations between sides of tiny triangulations follows from Euler’s relation.

This direct application of the framework to triangulations using an arbitrary spanning tree forces us, as explained in [1], to consider the catalog $D_{m,k}$ of all planar triangulations with $m$ edges and one boundary cycle, which is divided into $k$ sides. The entropy of this class of triangulation with a boundary is however $\alpha = 2.17$ bits per face (plus a $\beta k \log m$ term to take into account the number of sides into which the boundary is divided). As a consequence, the additive part of the entropy leads to a representation which is compact, but succinct only for class of triangulations with a boundary (recall that the class of triangulations of the sphere has only entropy 1.62 bits per triangle).

In summary, one can expect the general framework to yield easily compact representations for mesh connectivity with bounded face degrees. However, as we shall see, more care is needed to choose the decomposition in order to produce a succinct representation.

Local adjacency queries in $O(1)$ time

It still remains to observe that the multi-level structure (described by graphs $G$ and $G'$, together with the information associated with tiny pieces $M_j$), does allow to perform efficiently some local adjacency queries (neighboring queries on elementary cells).

Lemma 4. Let us consider an object $M$ which is provided with a decomposition $[(M_1, \ldots, M_p); G; G']$ as defined above. If each of the tiny pieces $M_j$ is a planar connected map having all faces of constant degree and a boundary of arbitrary linear size $O(n_j)$, then it is possible to answer in $O(1)$ time whether two elements of $M$ (vertices or faces) are adjacent or not.

Proof. The idea is that elementary cells (faces, vertices, edges) are shared by at most one or two adjacent tiny pieces (because of the definition of graph $G$), except for a small set of multiple cells (shared by an arbitrary number of tiny pieces). The planarity of graph $G$ ensures that the number of multiple cells is $O(\frac{n}{\log n})$, hence adjacency queries involving multiple cells can be answered with an additional information, which requires in overall a negligible amount of extra storage. Intuitively, local queries involving only a tiny piece $M_i$ are answered looking at the information contained in the corresponding explicit representation (stored in Table A), allowing $O(1)$ time navigation. Adjacency queries concerning elementary cells incident to the boundary of a tiny piece $M_j$, rely instead on the information in graphs $G$ and possibly $G'$. Some care must be taken to deal with the fact that a given cell at the meeting point between a lot of tiny pieces can have a non constant number of representations: however the number of such special cells is negligible and they can be detected at the $G'$ level. \qed
Attaching external data to elementary cells

Since our framework is designed to represent geometric objects (graphs, surfaces meshes, ...) it should be natural to allow dealing with other type of information\(^2\), such as vertex coordinates or other properties associated to elementary cells (color of faces, normals, ...). We can solve this problem by enriching our representation adding auxiliary information to the nodes of graph \(G\). More precisely, we can add to a node \(N_j\) the list of geometric data associated to cells lying in the tiny piece \(M_j\). This list contains the data associated to the internal cells of \(M_j\) and a selection of cells lying on its boundary, in such a way that boundary cells shared by two adjacent tiny pieces (not multiple cells) are stored only once. Multiple cells (shared by several tiny pieces) can have multiple copies: this does not affect the size of our representation, since their number is negligible. It suffices to list the set of multiple cells shared by tiny pieces: we distinguish between multiple cells relative to a same node \(N_j\) and those which are shared by tiny pieces corresponding to different nodes of \(G'\). There are very few cells of second type: since their number is \(\Theta(\frac{n}{\lg^2 n})\) storing their external data requires negligible auxiliary space. Concerning cells of first type, because our regrouping of nodes of \(G\), the number of such cells relative to a node \(N_j\) is at most \(O(\lg^2 n)\) (hence a reference costs \(O(\lg lg n)\) bits): then it suffices to store, for each node \(N_j\), the list of geometric data of a selection of these multiple cells, such that external data are stored only once. The same argument used before for the succinctness of the representation of \(G\) allows to guarantee the the overall number of copies concerning multiple cells does not affect the storage requirements of our encoding.

**Construction of the representation in \(O(n)\) time**

For concluding the presentation of our general framework we have to mention that we need further assumptions to allow \(O(n)\) time construction. Given an object \(M \in C_n\) of size \(n\), we assume we are able to compute a binary encoding for \(M\) of length at most \(O(\lg |C_n|)\) (in \(O(n)\) time). We need this assumption to guarantee that all references used in our representation are of size \(\Theta(\lg n)\); recall that we adopted the word – RAM model, where words of size \(O(\lg n)\) can be handled in \(O(1)\) time.

### 3 Representing 3-connected planar graphs

In this section we describe a particular catalog of tiny quadrangulations and an algorithm to decompose any irreducible quadrangulations into tiny regions taken from this catalog. This catalog satisfies Equation 1 so that the framework yields a succinct representation. Since irreducible quadrangulations are just another representation of 3-connected planar graphs, the result holds for these maps as well.

\(^2\)We suppose that external data can be stored on \(O(\lg n)\) bits.
Figure 1: First Pictures describe the closure defining the correspondence between plane rooted binary trees and irreducible dissections of the hexagon. The dissection and the corresponding 3-connected graph are also shown. Black circles (resp. white circles) represent vertices of the primal (dual) graph.

**Preliminaries on 3-connected graphs, quadrangulations and trees**

Plane trees are planar maps with only one face, the outer one. In other terms, plane trees only differ from the classical ordered trees in the fact that they are not rooted. As done in [10] we consider a special class of plane trees, binary trees where each vertex has 3 neighbors (and hence 2 sons if rooted). A vertex is a leaf if it has degree 1, otherwise it is an internal node.

Edges incident to leaves are called *stems*, remaining edges are called *inner edges*. A quadrangulation is a planar map having all its faces of degree 4. A dissection of the hexagon by quadrangular faces is a planar map whose outer face has degree 6 and inner faces have degree 4. A quadrangulation or dissection of the hexagon by quadrangular faces is said irreducible if it has no separating 4-cycle (we also call them *irreducible dissections*).

The following construction is more or less a folklore variation on the standard duality construction (also known as the Tutte’s bijection, see [17]) for planar maps: given a 3-connected planar map $M$, color its vertices in black and put a white vertex in the middle of each face and triangulate each face from this white vertex. The resulting new edges form a quadrangulation $Q$ (each face is made of the two triangles incident to an edge of $M$). It is not difficult to check that the 3-connectedness of $M$ is equivalent to the irreducibility of $Q$. Finally let us observe that it is possible to associate a rooted dissection $d$ of the hexagon to a rooted irreducible quadrangulation $Q$ by deleting the root edge.

Intuitively, a quadrangulation $Q$ can be viewed as an implicit representation of a planar map, induced with a bicoloration of its vertices: white (resp. black) vertices of $Q$ stand for faces (resp. vertices) of the original (resp. dual) map (see Fig. 1). More precisely, testing adjacency relations between vertices and faces in map $M$ corresponds to answering neighboring queries on vertices incident to the same face in the quadrangulation $Q$. It will prove convenient to describe our construction in terms of quadrangulations.
3.1 How does our framework apply

Our decomposition strategy will take advantage of the fact that irreducible dissections admit a special class of canonical spanning trees, described in [10]:

**Proposition 5.** There exists a bijection between the class of binary trees on $n$ internal nodes and the class of irreducible dissections with $n$ internal vertices, which can be computed in $O(n)$ time.

Let us suppose to have an irreducible dissection $Q$ of the hexagon with quadrangular faces. Following the traversal strategy explained in [10], it is possible to perform an opening algorithm on $Q$ which returns a binary tree: the result is a vertex spanning tree of $Q$, whose complete closure [10, Lemma 2] is exactly the original dissection of the hexagon (corresponding to a 3-connected planar graph). More precisely it is a binary tree $B$ on $n$ nodes having $n + 2$ stems and $(n - 1)$ inner edges (the correspondence is illustrated in Figure 1).

Let us recall a previous result concerning tree decompositions, stated in the following Lemma[19]:

**Lemma 6.** Given a binary tree $B$ on $n$ nodes and a positive integer parameter $\delta$, we can produce in linear time a partition of $B$ into a family of sub-trees $B_j$, whose sizes satisfy $\delta \leq \|B_j\| \leq 3\delta$.

Applying several times the algorithm above, we obtain a partition of $B$ into small binary trees having between $\frac{1}{3} \log^2 n$ and $\log^2 n$ nodes, which are then decomposed into tiny binary trees having between $\frac{1}{30} \log n$ and $\frac{1}{10} \log n$ nodes. Such a decomposition forms a partition of the edges of $B$ (which are edges of the original quadrangulation). Let us observe that only nodes, which are roots of tiny (resp. small) trees, are shared by different tiny (resp. small) trees: each of these nodes (having degree $d$) is split into a degree $d - 2$ root of one tree (descendent tree) and a leaf of another tree (ancestor tree). The way we have partitioned $B$ guarantees that number of tiny (resp. small) trees is $\Theta(\frac{n}{\log n})$ (resp. $\Theta(\frac{n}{\log^2 n})$).

3.2 Decomposition into tiny quadrangulations

The aim of this section is to describe a canonical way of obtaining a decomposition $\{M_1, \ldots, M_p\}$ of $Q$, starting from the decomposition of the vertex spanning tree $B$. Here we suppose we are given one tiny tree, denoted $TB_j$, having $n_j \leq \frac{1}{10} \log n$ nodes and $w_j$ false stems, obtained by decomposing $B$: we are going to show how to produce a tiny quadrangular map of size $\Theta(\log n)$ from $TB_j$. A tiny tree $TB_j$ has $n_j$ nodes and $n_j + 2$ stems: we now distinguish two kinds of leaves. Firstly there may exist some leaves which were already existing, as leaves, in the original tree $B$. On the other hand, there may exist some new leaves which correspond to internal nodes of $B$, before performing the decomposition into tiny trees: their incident edges are now called false stems. Recall that tiny trees are rooted (roots are shared by different tiny trees), and the number of false stems in a tiny tree is not fixed and it ranges in $0 \ldots n_j + 2$, depending on the way $B$ has been partitioned.
3.3 Local closure

Now it suffices to apply a "local closure" algorithm, inspired by the one introduced in [10], whose main steps are listed below (recall that in the original algorithm of [10], there was no distinction between true and false stems).

Let us perform a counterclockwise (ccw) traversal of the contour of $T_B^j$, starting from its root and walking along its edges (inner edges, stems and false edges).

- When traversing a true stem, which is preceded by 3 internal nodes (and not stems), its local closure consists in linking its incident node, with the preceding third node (on the boundary of the outer face) to create a quadrangular face (a white face in Figure 2).

- When traversing a (true) stem $s$, not preceded by 3 true nodes (original nodes existing in $T_B^j$), its local closure consists in attaching a dummy quadrangular face (a grey face in Fig. 2) to the boundary of $T_Q^j$, in such a way that the dummy face is then incident to the stem $s$ and does not enclose a false stem nor a dummy edge (i.e. an edge incident to a dummy face) previously added. Similarly, vertices incident to dummy faces, and not existing in the tiny tree, are called dummy vertices. Let us observe that we do not perform the merging of false stems (drawn with small circles in Figure 2).

In this way we produce a planar map whose internal faces are all quadrangles, and having an outer face of arbitrary size: inner edges (in the tree) may now be incident twice to the outer face.

Figure 2: Our local closure. The result of the closure operation does depend on the distribution of false stems. In our example, the binary tree has 11 inner vertices, 11 + 2 leaves (true and false stems, including the stem incident to the root), and can be optimally encoded by the binary word of length $2 \cdot 11$: 1101001011001001011000; while the corresponding false stems distribution is defined by the bit-vector of length 13 and weight 4: 000100001101.
3.4 Catalog of micro quadrangulations

The planar connected maps obtained in this way are also called tiny quadrangulations (the result of the procedure above is shown in Figure 2). Here we consider the catalog $\mathcal{D} = \{D_{p,k,w}\}$ containing all different quadrangulations obtained applying the strategy above. More precisely, an object $M_j$ of $\{D_{p,k,w}\}$ is a tiny quadrangulation $TQ$ obtained via our local closure from a tiny tree $TB$ having $p$ nodes, $p + 2$ stems and $wj$ false stems (the size parameter is the number $p$ of nodes in the tree). Moreover $TQ$ is induced with a partition of its boundary edges into $k$ sides. One can easily see that given a tiny binary tree with $p$ nodes, the corresponding tiny quadrangulation has a boundary of size at most $4p + 6$. \(^3\) We can then state the following bound on the size of catalog $\mathcal{D}$:

$$\log |D_{p,k,w}| \leq \log \left(2^p \cdot \binom{4p + 6}{k} \cdot \binom{p + 2}{w}\right) \approx 2p + k \log p + w \log p + O(k),$$

(since there are $\binom{p+2}{w}$ ways of distributing false stems, and $\binom{4p+6}{k}$ ways of partitioning boundary edges into sides; we have also $w \leq k$). The storage of elements of $\mathcal{D}$ (with their explicit representations) requires a negligible amount of space: recall that $p$, the number of nodes in a tiny tree, is at most $\frac{1}{10} \log n$.

Distribution of false stems

Once the initial spanning tree $B$ has been decomposed, we have to specify a rule for assigning (true) stems incident to nodes shared by tiny trees. We proceed as follows, assuming that $v$ is shared by two tiny trees $TB_j'$ and $TB_j''$, and denoting the possibly incident stem by $s$ (see Pictures in Fig. 3):

\(^3\)Concerning inner edges, doubly incident to the outer face, their number is $p-1$, hence their contribution to the boundary size is at most $2(mp - 1)$. On the other hand, the contribution of dummy faces is maximum when a dummy face is produced by a true stem immediately preceded by a false stem: as there are $\binom{p+2}{w}$ false stems, there may be at most $\lceil \frac{p+2}{2} \rceil$ dummy faces, each contributing for 4 edges. Finally the size of the boundary of a tiny quadrangulation is bounded by $2(p - 1) + 4(\frac{p+2}{2} + 1) = 4p + 6$.  

Figure 3: These Pictures explain how distributing stems and (dummy) faces between tiny quadrangulations sharing a node $v$ (a multiple node, represented with a small circle).
- if there exists in the descendant tree $B$ an inner edge $e$ incident to $v$ to the left of $s$, then we attach the stem $s$ to the tiny sub-tree $TB_j$, having $v$ as root and containing $e$;
- otherwise $s$ is the leftmost sibling of node $v$ in $B$, and we do attach $s$ to the ancestor tree $TB_j$. It is easy to observe that all the faces of the initial quadrangulation have been assigned to exactly one tiny quadrangulation, as they are incident each to one true stem.

3.5 Verification of the Hypotheses

Lemma 7. Given a quadrangulation $Q$ with $n$ vertices, our new splitting strategy produces a decomposition into tiny quadrangulations $\{TQ_1, \ldots, TQ_p\}$ satisfying Hypotheses 1, 2, 3, 4 and 5, and hence yields a succinct representation of $Q$ requiring asymptotically $2n + o(n)$ bits.

Proof. The additivity hypothesis holds, since (true) nodes in tiny quadrangulations (nodes of the tiny spanning tree) are not shared by tiny quadrangulations. The quadrangulation $Q$ is decomposed into tiny quadrangulations $TQ_j$ each containing between $\frac{1}{30} \log n$ and $\frac{1}{10} \log n$ nodes (nodes of the corresponding tiny vertex spanning tree): here the constant $c$ introduced in Section 2 is set to $\frac{1}{10}$. The graph $G$ used to describe adjacency relations between tiny quadrangulations is a planar map (each tiny quadrangulation is a connected planar map, whose edges may be incident twice to the outer face, and then doubly counted as boundary edges) having faces of degree at least 3 (for example, a degree 2 face incident to multiple edges can be contracted, observing that the corresponding sides are consecutive and shared by the same two tiny quadrangulations). Hence Euler's relation ensures that the number of arcs of $G$ (and hence the number of sides) is $O(\frac{n}{\log n})$. Hypotheses 1, 2, 3, 5 and 4 are satisfied by $\{TQ_1, \ldots, TQ_p\}$, hence Lemma 3 yields a succinct representation for the class of quadrangulations achieving the optimal asymptotic bound of 2 bits per vertex. A tiny colored quadrangulation is completely specified by: a Dyck word of length $2n_j$ (for the binary spanning tree), a binary word of length $n_j + 2$ and weight $w_j$ (describing false stems) and a binary word of length $4n_j + 6$ and weight $k_j$ (describing the partition of boundary edges of $TQ_j$ into sides). Hence $TQ_j$ can be encoded by a reference to an element in $D$, whose cost is $2n_j + w_j \log(n_j + 2) + k_j \log(4n_j + 6) \leq 2n_j + \beta k_j (\log n_j + O(1))$ (as $w_j \leq k_j$). The constant $c$ can be chosen so that the Table $A$ containing the explicit representations of the elements of $D$, requires $o(n)$ bits (recall that $n_j \leq \frac{1}{10} \log n$).

3.6 Representing 3-connected planar graphs

Given a 3-connected graph with $e$ edges, we first design a succinct representation of the associated quadrangulation. Since the corresponding vertex spanning tree has $e - 5$ inner nodes (vertices of the primal and dual graph) our representation requires asymptotically $2(e - 5) + o(e) = 2e + o(e)$ bits, according to Lemma 7. It
suffices to observe that testing adjacency between two vertices (resp. two faces) in a 3-connected graph, is equivalent to checking if two black (resp. white) nodes are opposite in the same quadrangular face, in the associated quadrangular map $Q$. This operation can be efficiently performed in $O(1)$ time with a slight modification of the standard adjacency queries considered in Lemma 4 (see Section 5 for more details).

### 3.7 Unique representations for vertices

As already observed, a number of elementary cells (multiple cells) are shared by several tiny pieces. One major problem to solve concerns the representation of vertices, which is not unique, as they can belong to more than one piece. One possible solution consists in exploiting the correspondence between vertices in the quadrangulation and nodes in the spanning tree. It simply suffices to associate to each vertex in $Q$ the corresponding node in the tree $B$.

Then inner nodes in a tiny tree $TB_j$ are uniquely specified. Remaining vertices, multiple vertices shared by tiny pieces, can be uniquely identified, saying that their canonical representative is the corresponding node in an adjacent tiny piece (in particular, for a multiple node shared by two tiny trees, its canonical representative is the leaf node in the ancestor tree).
Figure 5: These two Pictures illustrate the adjacency relations between tiny quadrangulations: since edges are allowed to be incident twice to the outer face, two tiny quadrangulations $TQ_j$ and $TQ_{j''}$ may be not adjacent even if "sharing" an edge $e$.

Figure 6: Vertex spanning tree bijection between rooted planar triangulations and rooted trees with 2 leaves per node.

4 Representing planar triangulations

Preliminaries on planar triangulations and trees

As done for 3-connected graphs, we take advantage of a recent bijection between planar triangulations and trees introduced in [20].

**Proposition 8.** There exists a 2$n$-to-2 correspondence between the class of rooted trees with $n$ nodes having 2 leaves per node, and the class of rooted planar triangulations with $n + 2$ vertices.

This correspondence relies on an opening/closure algorithm which computes a special vertex spanning tree (with two leaves per node) on $n$ nodes from a triangulation $T$, induced with its minimal realizer.

4.1 Decomposition of the graph

Here we suppose we are given a rooted triangulation $T$ with $n + 2$ vertices and the corresponding vertex spanning tree $B$ on $n$ nodes, whose complete closure is the initial triangulation (according to the closure/opening algorithm introduced
in [20]). Since there are no restrictions on the tree $B$ (it is just an ordered tree, with two leaves per node), we cannot in general decompose $B$ to get a partition into sub-trees as done for binary trees. Thus we describe below a different decomposition where few nodes can be shared by several sub-trees, but edges appears only once. First we distinguish in a subtree three kind of nodes: the leaves, the root and the internal nodes. Now we say that a family of sub-trees is a covering of $B$ if

— (i) each node of $B$ appears either in only one subtree as an internal node, either in several subtrees, in one subtree as a leave and others as the root.
— (ii) each edge of $B$ appears in exactly one subtree.
— (iii) the children of the root $r$ in one subtree correspond to consecutive children of $r$ as a node of $B$.

Then, the following Lemma controls the size of the subtrees. This Lemma is similar to the one designed for ordered trees [11], whose decomposition provides a covering of the nodes but not a partition of the edges (some edges may not appear in any subtree) and for which condition (iii) is also not guaranteed.

**Lemma 9.** Given a tree $B$ on $n$ nodes and $\delta \geq 2$, we can compute a family of sub-trees that is a covering of $B$. Their sizes satisfy $B_j \leq 3\delta - 2$ (and $B_j \geq \delta$, if the $B_j$ does not contain the root of $B$). Such a decomposition can be computed in linear time.

**Proof.** The following algorithm will construct the covering within the claimed complexity bound. We traverse $B$ in postfix order and maintain a subtree $\overline{B}$ which is the not yet decomposed part of $B$. In a node $r$ we perform the following operations depending on the size of the subtree of $\overline{B}$ rooted at $r$:

— if the size is smaller than $M$ then the size is returned to the parent of $r$ (thus the parent is able to evaluate the size of its subtree)
— if the size is between $M$ and $3M - 2$, the subtree rooted at $r$ is returned as a subtree of the covering, in $\overline{B}$, $r$ is replaced by a single leaf and 1 is returned to the parent as size of the subtree.
— if the size is bigger than $3M - 2$, $r$ has $\kappa$ children $s_1, s_2 \ldots s_k$ ($\kappa \geq 4$) whose size of subtree are $\sigma_1, \sigma_2 \ldots \sigma_k$. For all $i$ we have $1 \leq \sigma_i < M$, we consider the smallest value $j$ such that $\tau = \sum_{i=1}^{j} \sigma_i \geq M$, since $\sum_{i=1}^{j-1} \sigma_i \leq M - 1$ and $\sigma_j \leq M - 1$ we have $\tau \leq 2M - 2$. Then $r, \sigma_1, \sigma_2 \ldots \sigma_j$ is returned as a subtree of the covering (of size $\leq 2M - 1$), $\sigma_1, \sigma_2 \ldots \sigma_j$ are removed as children of $r$ in $\overline{B}$.

The process is continued to gather children of $r$, at the last step we may finish with $s_1 \ldots s_k$ with a size $\sum_{i=1}^{k} \sigma_i < M$, then this children are grouped with those of the preceding step giving a maximal size of $2M - 2 + M - 1 + 1 = 3M - 2$ for this subtree. Finally, $r$ is replaced by a single leaf in $\overline{B}$ and 1 is returned to the parent as size of the subtree.

Applying several times Lemma 9 to $B$, we obtain a family of small sub-trees covering $B$ of size $\Theta(\log^2 n)$, which are then decomposed into tiny sub-trees having $\Theta(\log n)$ nodes. This family of sub-trees forms a cover of the nodes of $B$ such that two sub-trees can intersect only at their root.
4.2 Tiny trees and tiny triangulations

As done for quadrangulations, it is possible to define a local closure algorithm (inspired by the one described in [20]), providing a correspondence between tiny trees and tiny triangulations. Firstly, if a tiny tree \( TB_j \) has an inner node \( v \) of degree 3, which appears as root of one (or more) different tiny tree \( TB_{j'} \) (neighbor of \( TB_j \)), we set \( v \) as false degree 3 inner node (in this case \( v \) had degree more than 3 in \( B \)).

Distribution of stems

Observe that tiny trees could have less than two leaves per node. In order to make a tiny tree belong to the same class as \( B \), we perform some modifications on it. Original stems in \( B \) are duplicated and distributed between tiny trees \( TB_j \) so that each node has two leaves. The strategy is similar to the one adopted in Section 3 for quadrangulations: we have only to take care that micro trees in this case are not binary trees, and each node must have exactly two incident stems. Duplicated stems, called false stems, are assigned to the tiny trees sharing a same node \( v \), in such a way that the cyclic order of neighbors around \( v \) is respected. Observe that false inner nodes are incident to at least one false stem, at the exception of the root node of a tiny tree (see Figure 9). It is straightforward to observe that the number of false degree 3 inner nodes and false (duplicated) stems is in overall \( O(\frac{n}{\lg n}) \).

4.3 Closure of a micro tree and catalog of micro triangulations

Our local closure of a tiny tree \( TB_j \) consists in performing a ccw traversal along its edges (starting from its root, in ccw order). As done in Section 3, we add a (true or dummy) triangular face by performing the merging of true stems: this depends on the distribution of false stems and false inner nodes. More precisely, we merge a stem in the following manner:

- if a true stem \( s \) is preceded by 2 (true) inner nodes, we link its incident node to the second preceding node to create a triangular face;

- otherwise, if \( s \) is preceded (on the boundary of the outer face) by a vertex \( v \) which is not an inner node we add a dummy triangular face incident to \( s \) and \( v \) (a grey face in Figure 8).
Figure 8: This Figure shows a tiny triangulation obtained via our local closure from a tiny tree $TB$. In our example we start with a tiny tree having 5 nodes and 10 leaves: there are 3 false stems (small circles) and 3 false inner nodes (including the root). The stems distribution is described by a binary word of length 10 and weight 3: 0010100001.

The planar connected maps $TT_j$ we obtain in this way (whose internal faces are all triangles, and with an outer face of arbitrary size) are called tiny triangulations.

4.4 Verification of the Hypotheses

**Lemma 10.** Given a planar triangulation $T$ with $n + 2$ vertices, our new splitting strategy produces a decomposition $\{TT_1, \ldots, TT_p\}$ into tiny colored triangulations satisfying the Hypotheses 1, 2, 3, 4 and 5, and hence yields a succinct representation of $T$ requiring asymptotically $3.24n + o(n)$ bits.

**Proof.** Our arguments rely on the same remarks used in the proof of Lemma 7. The additivity Hypothesis holds, since tiny triangulations only share duplicated nodes (roots of tiny trees), whose number is $O\left(\frac{n}{\lg n}\right)$. Here an object $M_j = TT_j$ is a tiny triangulation: $TT_j$ is obtained via our local closure from a tiny tree $TB_j$ having $n_j$ nodes, $2n_j$ stems and $w_j$ false stems; its boundary edges are partitioned into $k_j$ sides. Hence a tiny triangulation is described by: a binary word of length $4n_j - 2$ and weight $n_j - 1$ (there are about $2^{3.24n_j}$ such words, see [20]), a binary word of length $2n_j$ and weight $w_j$ (for false stems) and a binary word of length $6n_j - 2$ and weight $k_j$ (a tiny triangulation has at most $6n_j - 2$ boundary edges). Again the constant $c$ can be chosen so that Catalog $D$ requires an asymptotic negligible amount of space. \qed

5 Efficient local queries

In this section we discuss the implementation of a few natural supplementary queries that can easily be supported by our structures. A vertex in a tiny quadrangulation $TQ_j$ can be specified by a triple $(N'_i, a, v)$: where $N'_i$ is a node of map $G'$, corresponding to the small piece $SQ_i$ to which it belongs, $a$ is a local pointer to the zone of memory related to node the $N_j$, and $v$ is the index of
Figure 9: In these Pictures is depicted our strategy for constructing a decomposition of the original triangulation $T$ into tiny triangulations. At first the vertex spanning tree of $T$ is decomposed into tiny trees. We perform our local closure algorithm on tiny trees, producing tiny triangulations. Each tiny triangulation is provided with a partition into sides of its boundary edges describing neighboring relations.

the vertex in the explicit representation, stored in Table $A$, corresponding to $TQ_j$. As already observed in [1], in our encoding vertices may be not uniquely represented, because they are shared by different tiny pieces. Analogously, representing (quadrangular or triangular) faces is done specifying a triple $(N_i, a, f)$: their representation is uniquely defined, as faces are not shared by tiny pieces.

**Adjacency queries between faces and vertices**

Next Lemma provides an useful tool for local navigation between faces of adjacent tiny pieces: its proof relies on arguments similar to the ones detailed in [1], which are still valid for arbitrary tiny pieces (having faces of arbitrary constant degree).

**Lemma 11.** Given an object $M$ and its decomposition into tiny (and small) pieces $[(M_1, \ldots, M_p); G; G']$, it is possible to answer in constant time the following local queries:

- **Neighbor($f,e$):** given a face $f$ and an incident edge $e$, it returns the face $f'$ adjacent to $f$ containing $e$.
- **Adjacent($v,w$):** says if two vertices $v$ and $w$ are adjacent.

**Unique representation for vertices**

It is possible to avoid the problem of having multiple representations of vertices, by adopting a local labelling scheme and distinguishing vertices into 3 categories: vertices internal to a tiny piece, vertices shared by more than 2 tiny pieces, and vertices shared by more than 2 small pieces. As observed in Section 3.2, we can associate to each vertex a unique representation in a canonical way, using the correspondence between vertices in the original graph (quadrangulation or triangulation) and nodes in the vertex spanning tree. Next Lemma allows to deal with multiple vertices:
Lemma 12. It is possible to answer in $O(1)$ time the following queries involving multiple vertices, using asymptotically $o(n)$ extra bits:

- **Same**($\langle N_i', a, v \rangle, \langle N_i', a', w \rangle$): says if $v$ and $w$ represent the same vertex in the graph;
- **Node**($\langle N_i', a, v \rangle$): returns the canonical representative of vertex $v$ (a triple indicating the tiny and small sub-pieces to which $v$ belongs as node).

**Local adjacency queries on the quadrangulation**

Concerning the local navigation in the quadrangulation, our representation allows to perform in $O(1)$ time the following operation (which provides efficient navigation in the original 3-connected graph, as discussed in Section 3.6).

Lemma 13. In a quadrangulation $Q$ it is possible to answer in $O(1)$ time the following neighboring query:

- **Opposite**($v, w$): returns true if two vertices $v$ and $w$ are opposite and incident to the same quadrangular face in $Q$.

Proof. The validity of our arguments relies on the following properties that hold for 3-connected planar graphs and corresponding quadrangular irreducible dissections.

- The vertex spanning tree introduced in [10] is a binary tree, implying that the degree of nodes and the number of stems per node is bounded and constant.
- Every quadrangular face in the decomposition of $Q$ belongs exactly to one tiny quadrangulation. Moreover every node $v$ is incident to at most two dummy faces in the same tiny quadrangulation: for nodes with two stems, we observe that each dummy face is created by merging one of its stems; for nodes with one stem (possibly producing one dummy face), only one more dummy face may exist, possibly incident to the descendant node in the tree (see node $w$ in Figure 10).
- For each pair of vertices $v$ and $w$ there is at most one quadrangular face (if it exists) containing the two vertices and for which $v$ and $w$ are opposite (because the quadrangulation $Q$ has no separating 4-cycle).

Let $q$ be the quadrangular face possibly incident $v$ and $w$, let us call $v', v'', \ldots$ (resp. $w', w'', \ldots$) their copies, and let $TQ_j, TQ'_j, TQ''_j, \ldots$ denote the tiny quadrangulations containing them. We may suppose, without loss of generality, that $v$ precedes $w$ on the boundary of the spanning tree $B$, which is ccw oriented. For the sake of clarity, let us assume that $v$ and $w$ are not multiple nodes, roots of different tiny trees (as in Figure 10). If $v$ and $w$ are vertices lying in the same tiny quadrangulation (hence $j = j'$) we can answer by simply looking at the explicit representation stored in Table $A$: as $v$ and $w$ are at distance 2 in $TQ_j$, answering this query requires storing $o(n)$ extra bits. If the vertices belong to different tiny quadrangulations (not necessarily adjacent tiny quadrangulations) we retrieve at first their canonical representatives: if $\text{Canonical}(v)$ and $\text{Canonical}(w)$ belong to the same tiny quadrangulation, we are in the previous case and we proceed as before. Otherwise we proceed as follows, assuming that $v$ is preceding $w$ in ccw order. If $v$ and $w$ would be opposite and incident to a face $q$, then $q$ should
belong to the same tiny quadrangulation containing $w$ and be incident to one of the edges of $TB_j$ (inner edge or stem) which is incident to $w$. Since the degree of nodes in $B$ is bounded we know that there exist at most 2 quadrangular faces satisfying the last condition, say $q_1$ and $q_2$ (grey faces in Figure 10). Let us now retrieve the two vertices $v''$ and $x''$, opposite to $w$, which are incident to $q_1$ and $q_2$ (using information stored in Table $A$). Finally it suffices to test whether the $\text{Canonical}(v'')$ and $\text{Canonical}(x'')$ do coincide with $v$ (canonical representative of the copies of $v$). For concluding, we observe that the case where $v$ (or $w$) is a multiple node can be dealt in a similar manner: we can repeat the steps above a constant number of times, since each multiple node is contained in at most two different tiny trees.

\[\Box\]

6 Concluding remarks

We have presented a general framework for describing succinct representations of planar maps. In the particular case of 3-connected graphs and triangulations, we propose moreover canonical decompositions which, combined with the general framework, yield encodings that achieve asymptotically the information-theoretic lower bound for the storage, while supporting efficiently standard local navigation operations. The generality of our arguments suggests that our framework could apply to other popular encoding schemes, getting compact representations of other classes of planar graphs. This should take advantage of the similarities between explicit spanning tree coding [14, 23] and region-growing approaches [21, 22] as discussed in [12]. One interesting open problem is to extend optimal encodings [20] and compact representations of graphs [9, 8] to the case of higher genus triangulated surfaces. Although our results here are mainly theoretical (as in the case of previous works on compact representations [8, 9, 18]) this work has been a good source of inspiration for a practical
solution [4]. The main idea is that $n$ could be considered more as a constant than a parameter: so our aim is to design a hierarchical structure based on a splitting in tiny and small triangulations whose sizes are limited by finely tuned constants.

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