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# Lexicographic optimal homologous chains and applications to point cloud triangulations

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7 Abstract This paper considers a particular case of the Optimal Homologous Chain

 $_{\ensuremath{\scriptscriptstyle 8}}$  Problem (OHCP) for integer modulo 2 coefficients, where optimality is meant as a

<sup>9</sup> minimal lexicographic order on chains induced by a total order on simplices. The

<sup>10</sup> matrix reduction algorithm used for persistent homology is used to derive polynomial

algorithms solving this problem instance, whereas OHCP is NP-hard in the general

case. The complexity is further improved to a quasilinear algorithm by leveraging
 a dual graph minimum cut formulation when the simplicial complex is a pseudo-

a dual graph minimum cut formulation when the simplicial complex is a pseudomanifold. We then show how this particular instance of the problem is relevant, by

<sup>15</sup> providing an application in the context of point cloud triangulation.

<sup>16</sup> **Keywords** Simplicial homology, Point cloud triangulation, Delaunay triangulation

17 Mathematics Subject Classification (2010) 68U05

#### 18 1 Introduction

- 19 1.1 Problem statement
- <sup>20</sup> The computation of minimal simplicial homology generators has been a wide subject
- <sup>21</sup> of interest for its numerous applications related to shape analysis, computer graphics

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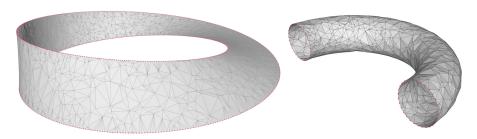


Fig. 1 Open surface triangulations under imposed boundaries (red cycles).

or computer-aided design. Coined in [21], we recall the Optimal Homologous Chain
 Problem (OHCP):

Problem 1 (OHCP) Given a *d*-chain *A* in a simplicial complex *K* and a set of weights given by a diagonal matrix *W* of appropriate dimension, find the  $L^1$ -norm minimal

<sup>26</sup> chain  $\Gamma_{\min}$  homologous to *A*:

<sup>27</sup>  $\Gamma_{\min} = \min_{\Gamma, B} ||W \cdot \Gamma||_1$  such that  $\Gamma = A + \partial_{d+1}B$  and  $\Gamma \in C_d(K)$ ,  $B \in C_{d+1}(K)$ 

<sup>28</sup> It has been shown that OHCP is NP-hard in the general case when using coefficients in

<sup>29</sup>  $\mathbb{Z}_2$  [9, 15]. However, we consider a specialization of this problem: the Lexicographic <sup>30</sup> Optimal Homologous Chain Problem (Lex-OHCP). Using coefficients in  $\mathbb{Z}_2$ , mini-<sup>31</sup> mality is now meant according to a lexicographic order on chains induced by a total <sup>32</sup> order on simplices. Formulated in the context of OHCP, this would require ordering <sup>33</sup> the simplices using a total order and taking a weight matrix *W* with generic term <sup>34</sup>  $W_{ii} = 2^i$ , where *i* is the rank along the total order, allowing the *L*<sup>1</sup>-norm minimization

<sup>35</sup> to be equivalent to a minimization along the lexicographic order.

### 36 1.2 Contributions

After providing some required definitions and notations (Section 2), we show how 37 an algorithm based on the matrix reduction algorithm used for the computation 38 of persistent homology [25] allows to solve this particular instance of OHCP in 39  $O(n^3)$  worst case complexity (Section 3). Using a very similar process, we show that 40 the problem of finding a minimal d-chain bounding a given (d - 1)-cycle admits 41 a similar algorithm with the same algorithmic complexity (Section 4). Section 5 42 then considers Lex-OHCP in the case where the simplicial complex K is a strongly 43 connected (d + 1)-pseudomanifold. By formulating it as a Lexicographic Minimum 44 Cut (LMC) dual problem, the algorithm can be improved to a quasilinear complexity: 45 the cost of sorting the dual edges and performing a  $O(E\alpha(E))$  algorithm based on 46 disjoint-sets, where E is the number of dual edges and  $\alpha$  is the inverse Ackermann 47 function. Finally, Section 6 legitimizes this restriction of OHCP by characterizing 48 the quality of lexicographic optimal homologous chains, namely in the context of 49 point cloud triangulation. After defining a total order closely related to the Delaunay 50

<sup>51</sup> triangulation, we provide details on an open surface algorithm given a boundary as

well as a watertight surface reconstruction algorithm given an interior and exterior information.

54 1.3 Related work

Several authors have studied algorithm complexities for the computation of  $L^1$ -norm 55 optimal cycles in homology classes [9, 10, 13-15, 21-23, 27, 37]. However, to the 56 best of our knowledge, considering lexicographic-minimal chains in their homology 57 classes is a new idea. When minimal cycles are of codimension 1 in a pseudo-58 manifold, the idea of considering the minimal cut problem on the dual graph has 59 been previously studied [35]. In particular, Chambers et al. [9] have considered graph 60 duality to derive complexity results for the computation of optimal homologous cycles 61 on 2-manifolds. Chen et al. [15] also use a reduction to a minimum cut problem 62 on a dual graph to compute minimal non-null homologous cycles on d-complexes 63 embedded in  $\mathbb{R}^d$ . Their polynomial algorithm (Theorem 5.2.3 in [15]) for computing 64 a homology class representative of minimal radius is reminiscent of our algorithm 65 for computing lexicographic minimal representatives (Algorithm 4). In a recent work 66 [22], Dey et al. consider the dual graph of pseudo-manifolds in order to obtain 67 polynomial time algorithms for computing minimal persistent cycles. Since they 68 consider arbitrary weights, they obtain the  $O(n^2)$  complexity of best known minimum 69 cut/maximum flow algorithms [34]. The lexicographic order introduced in our work 70 can be derived from the idea of a variational formulation of the Delaunay triangulation, 71 first introduced in [16] and further studied in [1, 17]. Finally, many methods have been 72 proposed to answer the problem of surface reconstruction in specific acquisition 73 contexts [30, 31, 33]: [32] classifies a large number of these methods according to the 74 assumptions and information used in addition to geometry. In the family of purely 75 geometric reconstruction based on a Delaunay triangulation, one early contribution 76 is the sculpting algorithm by Boissonnat [6]. The crust algorithm by Amenta et 77 al. [2, 3] and an algorithm based on natural neighbors [7] were the first algorithms 78 to guarantee a triangulation of the manifold under sampling conditions. However, 79 these general approaches usually have difficulties far from these sampling conditions, 80 in applications where point clouds are noisy or under-sampled. This difficulty can 81 be circumvented by providing additional information on the nature of the surface 82 [8, 20, 24]. Our contribution lies in this category of approaches. We provide some 83 topological information of the surface: a boundary for the open surface reconstruction 84 and interior and exterior regions for the closed surface reconstruction. 85

## 86 **2** Definitions

87 2.1 Simplicial complexes

<sup>88</sup> Consider an independent family  $A = (a_0, ..., a_d)$  of points of  $\mathbb{R}^N$ . We call a *d*-<sup>89</sup> simplex  $\sigma$  spanned by A the set of all points:

<sup>90</sup> 
$$x = \sum_{i=0}^{d} t_i a_i$$
, where  $\forall i \in \{0...d\}, t_i \ge 0$  and  $\sum_{i=0}^{d} t_i = 1$ 

Any simplex spanned by a subset of A is called a **face** of  $\sigma$ .

A simplicial complex K is a collection of simplices such that every face of a simplex of K is in K and the intersection of two simplices of K is either empty or a common face.

95 2.2 Simplicial chains

Let K be a simplicial complex of dimension at least d. The notion of chains can be

<sup>97</sup> defined with coefficients in any ring but we restrict here the definition to coefficients <sup>98</sup> in the field  $\mathbb{Z}_2 = \mathbb{Z}/2\mathbb{Z}$ . A *d*-chain *A* with coefficients in  $\mathbb{Z}_2$  is a formal sum of

<sup>99</sup> *d*-simplices :

100

$$A = \sum_{i} x_i \sigma_i, \text{ with } x_i \in \mathbb{Z}_2 \text{ and } \sigma_i \in K$$

We denote  $C_d(K)$  the vector space over the field  $\mathbb{Z}_2$  of *d*-chains in the complex *K*. Interpreting the coefficient  $x_i \in \mathbb{Z}_2 = \{0, 1\}$  in front of simplex  $\sigma_i$  as indicating the existence of  $\sigma_i$  in the chain *A*, we can view the *d*-chain *A* as a set of simplices : for a *d*-simplex  $\sigma$  and a *d*-chain *A*, we write that  $\sigma \in A$  if the coefficient for  $\sigma$  in *A* is 1. With this convention, the sum of two chains corresponds to the symmetric difference on their sets. In what follows, a *d*-simplex  $\sigma$  can also be interpreted as the *d*-chain containing only the *d*-simplex  $\sigma$ .

#### 108 2.3 Boundary operator

For a *d*-simplex  $\sigma = [a_0, \ldots, a_d]$ , the **boundary operator** is defined as the operator:

$$\partial_d : \boldsymbol{C_d} (K) \to \boldsymbol{C_{d-1}} (K)$$
$$\partial_d \sigma \stackrel{=}{\underset{\text{def.}}{=}} \sum_{i=0}^d [a_0, \dots, \widehat{a_i}, \dots, a_d]$$

where the symbol  $\hat{a}_i$  means the vertex  $a_i$  is deleted from the array. The kernel of the

- <sup>110</sup> boundary operator  $Z_d = \text{Ker } \partial_d$  is called the group of *d*-cycles and the image of the
- operator  $B_d = \text{Im} \partial_{p+1}$  is the group of d-boundaries. We say two d-chains A and A'
- are **homologous** if  $A A' = \partial_{d+1}B$ , for some (d + 1)-chain B.

#### 113 2.4 Lexicographic order

- We assume now a total order on the *d*-simplices of K,  $\sigma_1 < \cdots < \sigma_n$ , where
- <sup>115</sup>  $n = \dim C_d(K)$ . From this order, we define a lexicographic total order on *d*-chains.
- Definition 1 (Lexicographic total order) For  $C_1, C_2 \in C_d(K)$ :

117

$$C_1 \sqsubseteq_{lex} C_2 \iff \begin{cases} C_1 + C_2 = 0 \\ \text{or} \\ \sigma_{\max} = \max \{ \sigma \in C_1 + C_2 \} \in C_2 \end{cases}$$

This total order naturally extends to a strict total order  $\sqsubset_{lex}$  on  $C_d(K)$ .

#### 119 3 Lexicographic optimal homologous chain

- 120 3.1 Problem statement
- In this section, we define a particular instance of OHCP (Problem 1), the Lexicographic
   Optimal Homologous Chain Problem (Lex-OHCP):
- **Problem 2** (Lex-OHCP) Given a simplicial complex *K* with a total order on the d-simplices and a *d*-chain  $A \in C_d(K)$ , find the unique chain  $\Gamma_{\min}$  defined by :

$$\Gamma_{\min} = \min_{\sqsubseteq_{lex}} \left\{ \Gamma \in C_d(K) \mid \exists B \in C_{d+1}(K), \Gamma - A = \partial_{d+1}B \right\}$$

- Definition 2 A *d*-chain  $A \in C_d(K)$  is said reducible if there is a *d*-chain  $\Gamma \in C_d(K)$
- (called **reduction**) and a (d + 1)-chain  $B \in C_{d+1}(K)$  such that:

$$\Gamma \sqsubset_{lex} A \quad \text{and} \quad \Gamma - A = \partial_{d+1} B$$

<sup>129</sup> If this property cannot be verified, the *d*-chain *A* is said **irreducible**. If *A* is reducible,

we call total reduction of A the unique irreducible reduction of A. If A is irreducible,
 A is said to be its own total reduction.

Problem 2 can be reformulated as finding the total reduction of A.

#### <sup>133</sup> 3.2 Boundary matrix reduction

With  $m = \dim C_d(K)$  and  $n = \dim C_{d+1}(K)$ , we now consider the *m*-by-*n* boundary matrix  $\partial_{d+1}$  with entries in  $\mathbb{Z}_2$ . We enforce that rows of the matrix are ordered according to a given strict total order on *d*-simplices  $\sigma_1 < \cdots < \sigma_m$ , where the *d*-simplex  $\sigma_i$ is the basis element corresponding to the *i*<sup>th</sup> row of the boundary matrix. The order of columns, corresponding to an order on (d + 1)-simplices, is not relevant for this section and can be chosen arbitrarily.

For a matrix R, the index of the lowest (i.e. closest to the bottom) non-zero coefficient of a non-zero column  $R_j$  is denoted by low(j), or sometimes  $low(R_j)$  when we want to explicit the considered matrix.

<sup>143</sup> Algorithm 1 is a slightly modified version of the boundary reduction algorithm <sup>144</sup> presented in [25]. Indeed, for our purpose, we do not need the boundary matrix <sup>145</sup> storing all the simplices of all dimensions and apply the algorithm to the sub-matrix <sup>146</sup>  $\partial_{d+1} : C_{d+1}(K) \to C_d(K)$ . One checks easily that Algorithm 1 has  $O(mn^2)$  time <sup>147</sup> complexity. We now introduce a few lemmas useful for solving Problem 2. We allow

Algorithm 1: Reduction algorithm for the  $\partial_{d+1}$  matrixInput: A boundary matrix  $\partial_{d+1}$ Output: A reduced boundary matrix RR =  $\partial_{d+1}$ for  $j \leftarrow 1$  to n dowhile  $R_j \neq 0$  and  $\exists j_0 < j$  with  $low(j_0) = low(j)$  do $| R_j \leftarrow R_j + R_{j_0}$ endend

147

<sup>148</sup> ourselves to consider each column  $R_j$  of the matrix R, formally an element of  $\mathbb{Z}_2^m$ , as

the corresponding *d*-chain in the basis  $(\sigma_1, \ldots, \sigma_m)$ .

Lemma 1 A d-chain A is reducible if and only if at least one of its d-simplices is
reducible.

*Proof* If there is a reducible *d*-simplex  $\sigma \in A$ , *A* is reducible by the *d*-chain  $A' = A - \sigma + Red(\sigma)$ , where  $Red(\sigma)$  is a reduction for  $\sigma$ .

We suppose *A* to be reducible. Let  $\Gamma \sqsubset_{lex} A$  be a reduction for *A* and *B* the (d + 1)chain such that  $\Gamma - A = \partial B$ . We denote  $\sigma_{max} = \max \{\sigma \in A - \Gamma\}$ . Note that  $\sigma_{max}$  is homologous to  $\Gamma - A + \sigma_{max}$ . The chain  $\Gamma - A + \sigma_{max}$  only contains simplices smaller than  $\sigma_{max}$ , by definition of the lexicographic order (Definition 1). We have thus shown that if *A* is reducible, it contains at least one simplex that is reducible.

Lemma 2 After matrix reduction (Algorithm 1), a non-zero column  $R_j \neq 0$  can be described as

$$R_j = \sigma_{\text{low}(j)} + \Gamma$$
, where  $\Gamma$  is a reduction for  $\sigma_{\text{low}(j)}$ .

*Proof* As all matrix operations performed on *R* by the reduction algorithm consist of sums of columns of  $\partial_{d+1}$ , each non-zero column  $R_j$  of *R* is in the image of  $\partial_{d+1}$ . Therefore, there exists a (d + 1)-chain *B* such that  $R_j = \sigma_{low(j)} + \Gamma = \partial_{d+1}B$ , which, is equivalent in  $\mathbb{Z}_2$  to  $\Gamma - \sigma_{low(j)} = \partial_{d+1}B$ . By definition of the low of a column, we also have immediately:  $\Gamma \sqsubset_{lex} \sigma_{low(j)}$ . For each non-zero column, the largest simplex is therefore reducible by the other *d*-simplices of the column.

Lemma 3 After matrix reduction (Algorithm 1), there is a one-to-one correspondence between the reducible d-simplices and non-zero columns of R:

157 
$$\sigma_i \in C_d(K)$$
 is reducible  $\iff \exists j \in \{1..n\}, R_j \neq 0 \text{ and } \log(j) = i$ 

6

Proof Lemma 2 shows immediately that the simplex corresponding to the lowest index of a non-zero column is reducible. Suppose now that a *d*-simplex  $\tilde{\sigma}$  is reducible and let  $\tilde{\Gamma}$  be a reduction of it:  $\tilde{\sigma} + \tilde{\Gamma} = \partial_{d+1}B$  and  $\tilde{\Gamma} \sqsubset_{lex} \tilde{\sigma}$ . Algorithm 1 realizes the matrix factorization  $R = \partial_{d+1} \cdot V$ , where matrix *V* is invertible [25]. It follows that Im  $R = \text{Im } \partial_{d+1}$ . Therefore, non-zero columns of *R* span Im  $\partial_{d+1}$  and since  $\tilde{\sigma} + \tilde{\Gamma} = \partial_{d+1}B \in \text{Im } \partial_{d+1}$ , there is a family  $(R_j)_{j \in J} = (\sigma_{\text{low}(j)}, \Gamma_j)_{j \in J}$  of columns of *R* such that :

170

$$\tilde{\sigma} + \tilde{\Gamma} = \sum_{j \in J} \sigma_{\text{low}(j)} + \Gamma_j$$

<sup>166</sup> Every  $\sigma_{\text{low}(j)}$  represents the largest simplex of a column, and  $\Gamma_j$  a reduction chain for <sup>167</sup>  $\sigma_{\text{low}(j)}$ . As observed in section VII.1 of [25], one can check that the low indexes in *R* <sup>168</sup> are unique after the reduction algorithm. Therefore, there is a  $j_{\text{max}} \in J$  such that for

all j in  $J \setminus \{j_{\max}\}$ , low $(j) < low(j_{\max})$ , which implies:

$$\sigma_{j_{\max}} = \max\left\{\sigma \in \sum_{j \in J} \sigma_{\mathrm{low}(j)} + \Gamma_j\right\} = \max\left\{\sigma \in \tilde{\sigma} + \tilde{\Gamma}\right\} = \tilde{\sigma}$$

We have shown that for the reducible simplex  $\tilde{\sigma}$ , there is a non-zero column  $R_{j_{\text{max}}}$  with  $\tilde{\sigma} = \sigma_{\text{low}(j_{\text{max}})}$  as its largest simplex.

#### 171 3.3 Total reduction algorithm

<sup>172</sup> Combining the three previous lemmas give the intuition on how to construct the total

reduction solving Problem 2: Lemma 1 allows to consider each simplex individually,

Lemma 3 characterizes the reducible nature of a simplex using the reduced boundary

<sup>175</sup> matrix and Lemma 2 describes a column of the reduction boundary matrix as a simplex

and its reduction. We now present Algorithm 2, referred to as the **total reduction** algorithm. For a *d*-chain  $\Gamma$ ,  $\Gamma[i] \in \mathbb{Z}_2$  denotes the coefficient of the  $i^{th}$  simplex in the chain  $\Gamma$ .

Algorithm 2: Total reduction algorithm Inputs: A *d*-chain  $\Gamma$ , the reduced boundary matrix *R* Output: The total reduction of  $\Gamma$ for  $i \leftarrow m$  to 1 do if  $\Gamma[i] \neq 0$  and  $\exists j \in \{1..n\}$  with low(j) = i in *R* then  $\mid \Gamma \leftarrow \Gamma + R_j$ end end

178

**Proposition 1** Algorithm 2 finds the total reduction of a given d-chain in  $O(m^2)$  time complexity.

- <sup>181</sup> *Proof* In Algorithm 2, let  $\Gamma_{i-1}$  be the value of the variable  $\Gamma$  after iteration *i*. Since
- the iteration counter *i* decreases from m to 1, the input and output of the algorithm are

- respectively  $\Gamma_m$  and  $\Gamma_0$ . At each iteration,  $\Gamma_{i-1}$  are either equal to  $\Gamma_i$  or  $\Gamma_i + R_j$ . Since
- <sup>184</sup>  $R_j \in \text{Im } \partial_{d+1}, \Gamma_{i-1}$  is in both cases homologous to  $\Gamma_i$ . Therefore,  $\Gamma_0$  is homologous to
- <sup>185</sup>  $\Gamma_m$ . We are left to show that  $\Gamma_0$  is irreducible. From Lemma 1, it is enough to check <sup>186</sup> that it does not contain any reducible simplex.
- Let  $\sigma_i$  be a reducible simplex and let us show that  $\sigma_i \notin \Gamma_0$ . Two possibilities may occur:
- <sup>189</sup> if  $\sigma_i \in \Gamma_i$ , then  $\Gamma_{i-1} = \Gamma_i + R_j$ . Since low(j) = i, we have  $\sigma_i \in R_j$  and therefore <sup>190</sup>  $\sigma_i \notin \Gamma_{i-1}$ .
- <sup>191</sup> if  $\sigma_i \notin \Gamma_i$ , then  $\Gamma_{i-1} = \Gamma_i$  and again  $\sigma_i \notin \Gamma_{i-1}$ .
- <sup>192</sup> Furthermore, from iterations i 1 to 1, the added columns  $R_j$  contain only simplices <sup>193</sup> smaller than  $\sigma_i$  and therefore  $\sigma_i \notin \Gamma_{i-1} \Rightarrow \sigma_i \notin \Gamma_0$ .

Observe that using an auxiliary array allows to compute the correspondence  $low(j) \rightarrow i$  in time O(1). The column addition nested inside the loop lead to a  $O(m^2)$  time complexity for Algorithm 2.

- It follows that Problem 2 can be solved in  $O(mn^2)$  time complexity, by applying
- <sup>195</sup> successively Algorithms 1 and 2, or in  $O(N^3)$  complexity if N is the size of the <sup>196</sup> simplicial complex.

#### <sup>197</sup> 4 Lexicographic-minimal chain under imposed boundary

<sup>198</sup> 4.1 Problem statement

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- <sup>199</sup> This section will study a variant of Lex-OHCP (Problem 3) in order to solve the
- subsequent problem of finding a minimal *d*-chain bounding a given (d 1)-cycle (Problem 4).
- Problem 3 Given a simplicial complex K with a total order on the *d*-simplices and a *d*-chain  $\Gamma_0 \in C_d(K)$ , find :

$$\Gamma_{\min} = \min_{\sqsubseteq_{lex}} \left\{ \Gamma \in \boldsymbol{C}_{\boldsymbol{d}} \left( K \right) \mid \partial_{\boldsymbol{d}} \Gamma = \partial_{\boldsymbol{d}} \Gamma_0 \right\}$$

**Problem 4** Given a simplicial complex *K* with a total order on the *d*-simplices and a (d-1)-cycle *A*, check if *A* is a boundary:

$$B_{A} \stackrel{=}{=} \left\{ \Gamma \in \boldsymbol{C}_{\boldsymbol{d}} \left( K \right) \mid \partial_{\boldsymbol{d}} \Gamma = A \right\} \neq \emptyset$$

<sup>208</sup> If it is the case, find the minimal *d*-chain  $\Gamma$  bounded by *A*:

$$\Gamma_{min} = \min_{\Box} B_A$$

In Problem 4, finding a representative  $\Gamma_0$  in the set  $B_A \neq \emptyset$  such that  $\partial_d \Gamma_0 = A$  is

<sup>211</sup> sufficient: we are then taken back to Problem 3 to find the minimal *d*-chain  $\Gamma_{\min}$  such <sup>212</sup> that  $\partial_d \Gamma_{\min} = \partial_d \Gamma_0 = A$ .

#### 4.2 Boundary reduction transformation matrix 213

As in Section 3, we will derive an algorithmic solution to Problem 3 from the result 214

of the boundary matrix reduction algorithm. Note that, unlike Section 3 that used the 215

 $\partial_{d+1}$  boundary operator, we are now considering  $\partial_d$ , meaning the given total order on 216

d-simplices applies to the greater dimension. An arbitrary order can be taken for the 217

(d-1)-simplices to build the matrix  $\partial_d$ . Indeed, if we see the performed reduction in 218 matrix notation as  $R = \partial_d \cdot V$ , the minimization steps in this section will be performed

- 219 on the transformation matrix V, whose rows do follow the given simplicial ordering. 220
- The number of zero columns of R is the dimension of  $Z_d = \text{Ker } \partial_d$  [25]. Let's denote 221

222

223

it by  $n^{\text{Ker}} = \dim(Z_d)$ . By selecting all columns in V corresponding to zero columns in R, we obtain the matrix  $V^{\text{Ker}}$ , whose columns  $V_1^{\text{Ker}}, \ldots, V_{n^{\text{Ker}}}^{\text{Ker}}$  form a basis of  $Z_d$ . We first show a useful property on the matrix  $V^{\text{Ker}}$ . Note that the low index for any 224

column in  $V^{\text{Ker}}$  is well defined, as V is invertible. 225

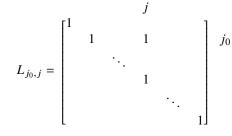
**Lemma 4** Indexes  $\{ low(V_i^{Ker}) \}_{i \in \{1..,n^{Ker}\}}$  are unique. If  $A \in Ker \partial_d \setminus \{0\}$ , there exists a unique column  $V_{max}^{Ker}$  of  $V^{Ker}$  with  $low(V_{max}^{Ker}) = low(A)$ . 226 227

*Proof* Before the boundary matrix reduction algorithm, the initial matrix V is the 228

identity: the low indexes are therefore unique. During iterations of the algorithm, the 229 matrix V is right-multiplied by an column-adding elementary matrix  $L_{i_0,i}$ , adding 230

column  $j_0$  to j with  $j_0 < j$ . 231





Therefore, the indexes  $\{low(V_i), V_i \in V\}$  stay on the diagonal (i.e.  $low(V_i) = i$ ) during 233 the reduction algorithm and are therefore unique. The restriction of V to  $V^{\text{Ker}}$  does 234 not change this property. 235

If  $A \in \text{Ker} \partial_d \setminus \{0\}$ , A can be written as a non-zero linear combination  $A = \sum_{i \in I} V_i^{\text{Ker}}$  of columns of  $V^{\text{Ker}}$ . By unicity of the lows of  $V^{\text{Ker}}$ , the largest low of the combination, i.e.  $\max_{i \in I} \{\text{low}(V_i^{\text{Ker}})\}$ , is in A and has to be the low of A.

- 4.3 Total reduction with imposed boundary 236
- We apply a similar total reduction algorithm as previously introduced in Section 3 but 237 using the matrix  $V^{\text{Ker}}$ . In the following algorithm,  $m = \dim C_d(K)$ . 238

**Proposition 2** Algorithm 3 computes the solution for Problem 3 in  $O(m^2)$  time com-239

plexity. 240

Algorithm 3: Total reduction variant

<b>Inputs:</b> A <i>d</i> -chain $\Gamma$ and $V^{\text{Ker}}$
<b>Output:</b> The minimal chain with the same boundary as $\Gamma$
for $i \leftarrow m$ to 1 do
<b>if</b> $\Gamma[i] \neq 0$ and $\exists j \in \{1n^{\text{Ker}}\}$ with $\text{low}(j) = i$ in $V^{\text{Ker}}$ then
<b>if</b> $\Gamma[i] \neq 0$ <b>and</b> $\exists j \in \{1n^{\text{Ker}}\}$ with $\text{low}(j) = i$ in $V^{\text{Ker}}$ <b>then</b> $  \Gamma \leftarrow \Gamma + V_j^{\text{Ker}}$
end
end

<sup>241</sup> *Proof* The proof is similar to the one of Proposition 1.

In Algorithm 3, we denote by  $\Gamma_{i-1}$  the value of variable  $\Gamma$  after iteration *i*. Since iteration counter *i* is decreasing from *m* to 1, the input and output of the algorithm are respectively  $\Gamma_m$  and  $\Gamma_0$ . Since  $V_j^{\text{Ker}} \in \text{Ker } \partial_d$ , at each iteration  $\partial\Gamma_{i-1} = \partial\Gamma_i$  therefore  $\partial\Gamma_0 = \partial\Gamma_m$ . We are left to show the algorithm's result is the minimal solution.

Suppose there is  $\Gamma^*$  such that  $\partial_d \Gamma^* = \partial \Gamma$  and  $\Gamma^* \sqsubset_{lex} \Gamma_0$ . Let's consider the

difference  $\Gamma_0 - \Gamma^*$ , and its largest element index low $(\Gamma_0 - \Gamma^*) = i$ , with  $\sigma_i \in \Gamma_0$ and  $\sigma_i \notin \Gamma^*$  by Definition 1 of the lexicographic order. As  $\Gamma_0 - \Gamma^* \in \text{Ker } \partial_d \setminus \{0\}$ ,

there has to be a column  $V_j^{\text{Ker}}$  in  $V^{\text{Ker}}$  where low  $\left(V_j^{\text{Ker}}\right) = i$ , from Lemma 4. Two

- possibilities may occur at iteration i:
- if  $\sigma_i \in \Gamma_i$ , then  $\Gamma_{i-1} = \Gamma_i + V_j^{\text{Ker}}$ . Since i = low(j), we have  $\sigma_i \in V_j^{\text{Ker}}$  and therefore  $\sigma_i \notin \Gamma_{i-1}$ .
- <sup>253</sup> − if  $\sigma_i \notin \Gamma_i$ , then  $\Gamma_{i-1} = \Gamma_i$  and again  $\sigma_i \notin \Gamma_{i-1}$ .

However, from iterations i-1 to 1, the added columns  $V_j^{\text{Ker}}$  contains only simplices with indices smaller than i and therefore we obtain  $\sigma_i \notin \Gamma_{i-1} \Rightarrow \sigma_i \notin \Gamma_0$ , a contradiction to the definition of  $\sigma_i$  as the largest element of  $\Gamma^* - \Gamma_0$ .

 $_{254}$  4.4 Finding a representative of **B**<sub>A</sub>

As previously mentioned, solving Problem 4 requires deciding if the set  $B_A$  is empty

and in case it is not empty, finding an element of the set  $B_A$ . Algorithm 3 can then be

used to minimize this element under imposed boundary. In the following algorithm,

<sup>258</sup>  $m = \dim C_{d-1}(K)$  and  $n = \dim C_d(K)$ .

**Proposition 3** Algorithm 4 decides if the set  $B_A$  is non-empty, and in that case, finds a representative  $\Gamma_0$  such that  $\partial\Gamma_0 = A$  in  $O(m^2)$  time complexity.

<sup>261</sup> *Proof* We start by two trivial observations from the definition of a reduction. First, A

is a boundary if and only if its total reduction is the null chain. Second, if a non-null
 chain is a boundary, then its greatest simplex is reducible.

- If, at iteration  $i, A_0[i] \neq 0$ , then  $\sigma_i$  is the greatest simplex in  $A_0$ . In the case R has
- no column  $R_i$  such that low(j) = i,  $\sigma_i$  is not reducible by Lemma 3 and therefore  $A_0$

is not a boundary. Since A and  $A_0$  differ by a boundary (added columns of R), A is

- not a boundary either. This means the set  $B_A$  is empty.
- <sup>266</sup> The main difference with the previous chain reduction is we keep track of the column

#### Algorithm 4: Finding a representative

Inputs: A (d-1)-chain A, a boundary matrix R reduced by VOutput: A d-chain  $\Gamma_0$  bounded by A if A is a boundary.  $\Gamma_0 \leftarrow \emptyset$   $A_0 \leftarrow A$ for  $i \leftarrow m$  to 1 do if  $A_0[i] \neq 0$  then if  $\exists j \in \{1...n\}$  with low(j) = i in R then  $\begin{vmatrix} A_0 \leftarrow A_0 + R_j \\ & & \\ & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & &$ 

- <sup>269</sup> operations in  $\Gamma_0$ . If the total reduction of A is null, we have found a linear combination <sup>270</sup>  $(R_i)_{i \in I}$  such that  $A = \sum_{i \in I} R_i$ . We have also computed  $\Gamma_0$  as the sum of the
- ( $R_j$ )<sub> $j \in J$ </sub> such that  $A = \sum_{j \in J} R_j$ . We have also computed  $\Gamma_0$  as the sum of the corresponding columns in V:  $\Gamma_0 = \sum_{j \in J} V_j$ . As  $R = \partial_d \cdot V$ , we can now verify:

$$\partial_d \Gamma_0 = \partial_d \left( \sum_{j \in J} V_j \right) = \sum_{j \in J} R_j = A$$

#### <sup>273</sup> 5 Efficient algorithm for codimension 1 (dual graph)

In this section we focus on Problem 5, a specialization of Problem 2, namely when K is a subcomplex of a (d + 1)-pseudomanifold.

#### 276 5.1 Problem statement

Recall that a *d*-dimensional simplicial complex is said **pure** if it is of dimension dand any simplex has at least one coface of dimension d.

**Definition 3** A *d*-pseudomanifold is a pure *d*-dimensional simplicial complex for which each (d - 1)-face has exactly two *d*-dimensional cofaces.

**Definition 4** The **dual graph** of a *d*-pseudomanifold  $\mathcal{M}$  is the graph whose vertices are in one-to-one correspondence with the *d*-simplices of  $\mathcal{M}$  and whose edges are in one-to-one correspondence with (d - 1)-simplices of  $\mathcal{M}$ : an edge *e* connects two vertices  $v_1$  and  $v_2$  of the graph if and only if *e* corresponds to the (d - 1)-face with cofaces corresponding to  $v_1$  and  $v_2$ .

Definition 5 A strongly connected *d*-pseudomanifold is a *d*-pseudomanifold whose
 dual graph is connected.

Given a strongly connected (d + 1)-pseudomanifold  $\mathcal{M}$  and  $\tau_1 \neq \tau_2$  two (d + 1)simplices of  $\mathcal{M}$ , we consider a special case of Problem 2 where  $K = \mathcal{M} \setminus {\tau_1, \tau_2}$  and  $A = \partial \tau_1$ :

Problem 5 Given a strongly connected (d + 1)-pseudomanifold  $\mathcal{M}$  with a total order on the *d*-simplices and two distinct (d + 1)-simplices  $(\tau_1, \tau_2)$  of  $\mathcal{M}$ , find:

<sup>293</sup> 
$$\Gamma_{\min} = \min_{\square_{lex}} \left\{ \Gamma \in C_d \left( \mathcal{M} \right) \mid \exists B \in C_{d+1} \left( \mathcal{M} \setminus \{\tau_1, \tau_2\} \right), \Gamma - \partial \tau_1 = \partial B \right\}$$

Definition 6 Seeing a graph G as a 1-dimensional simplicial complex, we define

the **coboundary operator**  $\partial^0 : C_0(G) \to C_1(G)$  as the linear operator defined by the transpose of the matrix of the boundary operator  $\partial_1 : C_1(G) \to C_0(G)$  in the canonical basis of simplices.<sup>1</sup>

For a given graph  $G = (\mathcal{V}, \mathcal{E})$ ,  $\mathcal{V}$  and  $\mathcal{E}$  respectively denote its vertex and edge sets. For a *d*-chain  $\alpha \in C_d(\mathcal{M})$  and a (d + 1)-chain  $\beta \in C_{d+1}(\mathcal{M})$ ,  $\tilde{\alpha}$  and  $\tilde{\beta}$  denote their

respective dual 1-chain and dual 0-chain in the dual graph  $G(\mathcal{M})$  of  $\mathcal{M}$ . We easily see that:

Remark 1 For a set of vertices  $\mathcal{V}_0 \subset \mathcal{V}, \partial^0 \mathcal{V}_0$  is exactly the set of edges in  $G = (\mathcal{V}, \mathcal{E})$ that connect vertices in  $\mathcal{V}_0$  with vertices in  $\mathcal{V} \setminus \mathcal{V}_0$ .

Remark 2 Let  $\mathcal{M}$  be a (d + 1)-pseudomanifold. If  $\alpha \in C_d(\mathcal{M})$  and  $\beta \in C_{d+1}(\mathcal{M})$ , then  $\tilde{\alpha} = \partial^0 \tilde{\beta} \iff \alpha = \partial_{d+1} \beta$ .

<sup>306</sup> 5.2 Codimension 1 and Lexicographic Min Cut (LMC)

The order on *d*-simplices of a (d + 1)-pseudomanifold  $\mathcal{M}$  naturally defines a corresponding order on the edges of the dual graph:  $\tau_1 < \tau_2 \iff \tilde{\tau}_1 < \tilde{\tau}_2$ . This order extends similarly to a lexicographic order  $\sqsubseteq_{lex}$  on sets of edges (or, equivalently, 1-chains) in the graph.

In what follows, we say a set of edges  $\tilde{\Gamma}$  **disconnects**  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  in the graph  $(\mathcal{V}, \mathcal{E})$ if  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  are not in the same connected component of the graph  $(\mathcal{V}, \mathcal{E} \setminus \tilde{\Gamma})$ .

Given a graph with weighted edges and two vertices, the min-cut/max-flow problem [26, 34] consists in finding the minimum cut (i.e. set of edges) disconnecting the two vertices, where minimum is meant as minimal sum of weights of cut edges. We consider a similar problem where the minimum is meant in term of a lexicographic order: the Lexicographic Min Cut (LMC).

Problem 6 (LMC) Given a connected graph  $G = (\mathcal{V}, \mathcal{E})$  with a total order on  $\mathcal{E}$  and

two vertices  $\tilde{\tau}_1, \tilde{\tau}_2 \in \mathcal{V}$ , find the set  $\tilde{\Gamma}_{LMC} \subset \mathcal{E}$  minimal for the lexicographic order

 $<sup>\</sup>sqsubseteq_{lex}$ , that disconnects  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  in G.

<sup>&</sup>lt;sup>1</sup> In order to avoid to introduce non essential formal definitions, the coboundary operator is defined over chains since, for finite simplicial complexes, the canonical inner product defines a natural bijection between chains and cochains.

Proposition 4  $\Gamma_{\min}$  is solution of Problem 5 if and if only its dual 1-chain  $\tilde{\Gamma}_{\min}$  is solution of Problem 6 on the dual graph  $G(\mathcal{M})$  of  $\mathcal{M}$  where  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  are respective

<sup>323</sup> *dual vertices of*  $\tau_1$  *and*  $\tau_2$ *.* 

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<sup>324</sup> *Proof* Problem 5 can be equivalently formulated as:

$$\Gamma_{\min} = \min_{\sqsubseteq_{lex}} \left\{ \partial_{d+1}(\tau_1 + B) \mid B \in C_{d+1}\left(\mathcal{M} \setminus \{\tau_1, \tau_2\}\right) \right\}$$
(1)

Using Observation 2, we see that  $\Gamma_{\min}$  is the minimum in Equation (1) if and only if its dual 1-chain  $\tilde{\Gamma}_{\min}$  satisfies:

$$\tilde{\Gamma}_{\min} = \min_{\Box_{lex}} \left\{ \partial^0 (\tilde{\tau}_1 + \tilde{B}) \mid \tilde{B} \subset \mathcal{V} \setminus \{\tilde{\tau}_1, \tilde{\tau}_2\} \right\}$$
(2)

- <sup>329</sup> Denoting  $\tilde{\Gamma}_{LMC}$  the minimum of Problem 6, we need to show that  $\tilde{\Gamma}_{LMC} = \tilde{\Gamma}_{min}$ .
- As  $\tilde{\Gamma}_{LMC}$  disconnects  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  in  $G = (\mathcal{V}, \mathcal{E})$ ,  $\tilde{\tau}_2$  is not in the connected component of
- $\tilde{\tau}_1$  in  $(\mathcal{V}, \mathcal{E} \setminus \tilde{\Gamma}_{LMC})$ . We define  $\tilde{B}$  as the connected component of  $\tilde{\tau}_1$  in  $(\mathcal{V}, \mathcal{E} \setminus \tilde{\Gamma}_{LMC})$
- minus  $\tilde{\tau}_1$ . We have that  $\tilde{B} \subset \mathcal{V} \setminus {\tilde{\tau}_1, \tilde{\tau}_2}$ . Consider an edge  $e \in \partial^0(\tilde{\tau}_1 + \tilde{B})$ . From
- Observation 1, *e* connects a vertex  $v_a \in {\tilde{\tau}_1} \cup \tilde{B}$  with a vertex  $v_b \notin {\tilde{\tau}_1} \cup \tilde{B}$ . From
- the definition of  $\tilde{B}$ ,  $\tilde{\Gamma}_{LMC}$  disconnects  $v_a$  and  $v_b$  in G, which in turn implies  $e \in \tilde{\Gamma}_{LMC}$ .
- We have therefore shown that  $\partial^0(\tilde{\tau}_1 + \tilde{B}) \subset \tilde{\Gamma}_{LMC}$ . Using Equation (2), we get:

$$\tilde{\Gamma}_{\min} \sqsubseteq_{lex} \partial^0 (\tilde{\tau}_1 + \tilde{B}) \sqsubseteq_{lex} \tilde{\Gamma}_{LMC}$$
(3)

Now we claim that if there is a  $\tilde{C} \subset \mathcal{V} \setminus {\tilde{\tau}_1, \tilde{\tau}_2}$  with  $\tilde{\Gamma} = \partial^0(\tilde{\tau}_1 + \tilde{C})$ , then  $\tilde{\Gamma}$  disconnects  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  in  $(\mathcal{V}, \mathcal{E})$ . Consider a path in *G* from  $\tilde{\tau}_1$  to  $\tilde{\tau}_2$ . Let  $v_a$  be the last vertex of the path that belongs to  ${\tilde{\tau}_1} \cup \tilde{C}$  and  $v_b$  the next vertex on the path (which exists since  $\tilde{\tau}_2$  is not in  ${\tilde{\tau}_1} \cup \tilde{C}$ ). From Observation 1, we see that the edge  $(v_a, v_b)$  must belong to  $\tilde{\Gamma} = \partial^0(\tilde{\tau}_1 + \tilde{C})$ . We have shown that any path in *G* connecting  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  has to contain an edge in  $\tilde{\Gamma}$  and the claim is proved.

In particular, the minimum  $\tilde{\Gamma}_{\min}$  disconnects  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  in  $(\mathcal{V}, \mathcal{E})$ . As  $\tilde{\Gamma}_{LMC}$  denotes the minimum of Problem 6,  $\tilde{\Gamma}_{LMC} \sqsubseteq_{lex} \tilde{\Gamma}_{\min}$  which, together with Equation (3), gives us  $\tilde{\Gamma}_{LMC} = \tilde{\Gamma}_{\min}$ . We have therefore shown the minimum defined by Equation (2) coincides with the minimum defined in Problem 6.

#### <sup>337</sup> 5.3 Algorithm for Lexicographic Min Cut

- <sup>338</sup> In light of the new problem equivalency, we will study an algorithm solving Problem
- 6. As we will only consider the dual graph for this section, we leave behind the dual

chain notation: vertices  $\tilde{\tau}_1$  and  $\tilde{\tau}_2$  are replaced by  $\alpha_1$  and  $\alpha_2$ , and the solution to the

- problem is simply noted  $\Gamma_{\text{LMC}}$ . The following lemma exposes a constructive property of the solution on subgraphs.
- Lemma 5 Consider  $\Gamma_{\text{LMC}}$  solution of Problem 6 for the graph  $G = (\mathcal{V}, \mathcal{E})$  and  $\alpha_1, \alpha_2 \in \mathcal{V}$ . Let  $e_0$  be an edge in  $\mathcal{V} \times \mathcal{V}$  such that  $e_0 < \min\{e \in \mathcal{E}\}$ . Then:
- The solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$  is either  $\Gamma_{\text{LMC}}$  or  $\Gamma_{\text{LMC}} \cup \{e_0\}$ .

-  $\Gamma_{\text{LMC}} \cup \{e_0\}$  is solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$  if and only if  $\alpha_1$  and  $\alpha_2$  are connected in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\} \setminus \Gamma_{\text{LMC}})$ .

<sup>348</sup> Proof Let's call  $\Gamma'_{LMC}$  the solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$ . Since  $\Gamma'_{LMC} \cap \mathcal{E}$  disconnects <sup>349</sup>  $\alpha_1$  and  $\alpha_2$  in  $(\mathcal{V}, \mathcal{E})$ , one has  $\Gamma_{LMC} \sqsubseteq_{lex} \Gamma'_{LMC}$ . Since  $\Gamma_{LMC} \cup \{e_0\}$  disconnects <sup>350</sup>  $\alpha_1$  and  $\alpha_2$  in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$ , we also have  $\Gamma'_{LMC} \sqsubseteq_{lex} \Gamma_{LMC} \cup \{e_0\}$ . Therefore,

<sup>351</sup>  $\Gamma_{\text{LMC}} \sqsubseteq_{lex} \Gamma'_{\text{LMC}} \sqsubseteq_{lex} \Gamma_{\text{LMC}} \cup \{e_0\}.$ 

As  $e_0 < \min\{e \in \mathcal{E}\}$ , there is no set in  $\mathcal{E} \cup \{e_0\}$  strictly between  $\Gamma_{\text{LMC}}$  and  $\Gamma_{\text{LMC}} \cup \{e_0\}$  for the lexicographic order. It follows that  $\Gamma'_{\text{LMC}}$  is either equal to  $\Gamma_{\text{LMC}}$  or  $\Gamma_{\text{LMC}} \cup \{e_0\}$ . The choice for  $\Gamma'_{\text{LMC}}$  is therefore ruled by the property that it should disconnect  $\alpha_1$  and  $\alpha_2$ : if  $\alpha_1$  and  $\alpha_2$  are connected in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\} \setminus \Gamma_{\text{LMC}})$ ,  $\Gamma_{\text{LMC}}$  does not disconnect  $\alpha_1$  and  $\alpha_2$  in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$  and  $\Gamma_{\text{LMC}} \cup \{e_0\}$  has to be the solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$ . On the other hand, if  $\alpha_1$  and  $\alpha_2$  are not connected in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\} \setminus \Gamma_{\text{LMC}})$ , then both  $\Gamma_{\text{LMC}}$  and  $\Gamma_{\text{LMC}} \cup \{e_0\}$  disconnect  $\alpha_1$  and  $\alpha_2$  in  $(\mathcal{V}, \mathcal{E} \cup \{e_0\} \setminus \Gamma_{\text{LMC}})$ , but as  $\Gamma_{\text{LMC}} \sqsubset_{lex} \Gamma_{\text{LMC}} \cup \{e_0\}$ ,  $\Gamma_{\text{LMC}} \cup \{e_0\}$  is not the solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$ , but as  $\Gamma_{\text{LMC}} \sqsubset_{lex} \Gamma_{\text{LMC}} \cup \{e_0\}$ ,  $\Gamma_{\text{LMC}} \cup \{e_0\}$  is not the solution for  $(\mathcal{V}, \mathcal{E} \cup \{e_0\})$ .

Building an algorithm from Lemma 5 suggests a data structure able to check 352 if vertices  $\alpha_1$  and  $\alpha_2$  are connected in the graph: the disjoint-set data structure, 353 introduced for finding connected components [28], does exactly that. In this structure, 354 each set of elements has a different root value, called representative. Calling the 355 operation MakeSet on an element creates a new set containing this element. The 356 FindSet operation, given an element of a set, returns the representative of the set. For 357 all elements of the same set, FindSet will of course return the same representative. 358 Finally, the structure allows merging two sets, by using the **UnionSet** operation. After 359 this operation, elements of both sets will have the same representative. 360

We now describe Algorithm 5. The algorithm expects a set of edges sorted in decreasing order according to the lexicographic order.

```
Algorithm 5: Lexicographic Min Cut
   Inputs: G = (\mathcal{V}, \mathcal{E}) and \alpha_1, \alpha_2 \in \mathcal{V}, with \mathcal{E} = \{e_i, i = 1, \dots, n\} in decreasing order
   Output: \Gamma_{LMC}
   \Gamma_{LMC} \leftarrow \emptyset
   for v \in \mathcal{V} do
        MakeSet(v)
   end
   for e \in \mathcal{E} in decreasing order do
          e = (v_1, v_2) \in \mathcal{V} \times \mathcal{V}
          r_1 \leftarrow \mathbf{FindSet}(v_1), r_2 \leftarrow \mathbf{FindSet}(v_2)
           c_1 \leftarrow \mathbf{FindSet}(\alpha_1), c_2 \leftarrow \mathbf{FindSet}(\alpha_2)
          if \{r_1, r_2\} = \{c_1, c_2\} then
                \Gamma_{\text{LMC}} \leftarrow \Gamma_{\text{LMC}} \cup e
            else
                  UnionSet(r_1, r_2)
            end
   end
```

**Proposition 5** Algorithm 5 computes the solution of Problem 6 for a given graph 363

 $(\mathcal{V}, \mathcal{E})$  and two vertices  $\alpha_1, \alpha_2 \in \mathcal{V}$ . Assuming the input set of edges  $\mathcal{E}$  are sorted, 364

the algorithm has  $O(n\alpha(n))$  time complexity, where n is the cardinal of  $\mathcal{E}$  and  $\alpha$  the 365

inverse Ackermann function. 366

*Proof* We denote by  $e_i$  the  $i^{th}$  edge along the decreasing order and  $\Gamma_{LMC}^i$  the value 367 of the variable  $\Gamma_{LMC}$  of the algorithm after iteration *i*. The algorithm works by 368 incrementally adding edges in decreasing order and tracking the growing connected 369 components of the set associated with  $\alpha_1$  and  $\alpha_2$  in  $(\mathcal{V}, \{e \in \mathcal{E}, e \ge e_i\} \setminus \Gamma_{IMC}^i)$ , for 370  $i=1,\ldots,n.$ 371

At the beginning, no edges are inserted, and  $\Gamma_{LMC}^0 = \emptyset$  is indeed solution for  $(\mathcal{V}, \emptyset)$ . With Lemma 5, we are guaranteed at each iteration *i* to find the solution for 372 373  $(\mathcal{V}, \{e \in \mathcal{E}, e \ge e_i\})$  by only adding to  $\Gamma_{\text{LMC}}^{i-1}$  the current edge  $e_i$  if  $\alpha_1$  and  $\alpha_2$  are 374 connected in  $\{e \in \mathcal{E}, e \ge e_i\} \setminus \Gamma_{\text{LMC}}^{i-1}$ , which is done in the if-statement. If the edge is 375 not added, we update the connectivity of the graph  $(\mathcal{V}, \{e \in \mathcal{E}, e \ge e_i\} \setminus \Gamma_{LMC}^i)$  by 376 merging the two sets represented by  $r_1$  and  $r_2$ . After each iteration,  $\Gamma_{LMC}^i$  is solution 377 for  $(\mathcal{V}, \{e \in \mathcal{E}, e \ge e_i\})$  and when all edges are processed,  $\Gamma_{LMC}^n$  is solution for 378  $(\mathcal{V}, \mathcal{E}).$ 379

The complexity of the MakeSet, FindSet and UnionSet operations have been shown to be respectively O(1),  $O(\alpha(v))$  and  $O(\alpha(v))$ , where  $\alpha(v)$  is the inverse Ackermann function on the cardinal of the vertex set [36]. Assuming sorted edges as input of the algorithm – which is performed in  $O(n \ln(n))$ , the algorithm runs in  $O(n\alpha(n))$  time complexity. П

The similarity of Algorithm 5 with Kruskal's algorithm for minimum spanning-380

tree suggests an even better theoretical time complexity, by using Chazelle's algorithm 381

[12] for minimum spanning-tree, running in  $O(n\alpha(n))$  complexity without requiring 382

sorted edges. 383

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#### 6 Application to point cloud triangulation 384

In all that precedes, the order on simplices was not specified and one can wonder if 385 choosing such an ordering makes the specialization of OHCP too restrictive for it to 386 be useful. In this section, we give a concrete example where this restriction makes 387 sense and provides a simple and elegant application to the problem of point cloud 388 triangulation. Whereas all that preceded dealt with an abstract simplicial complex, 389 we now consider a bijection between vertices and a set of points in Euclidean space, 390 allowing to compute geometric quantities on simplices.

6.1 Simplicial ordering

Recent works have studied a characterization of the 2D Delaunay triangulation as a 393

lexicographic minimum over 2-chains. Denote by  $R_B(\sigma)$  the radius of the smallest 39 enclosing ball and  $R_{\rm C}(\sigma)$  the radius of the circumcircle of a 2-simplex  $\sigma$ . For an 395



Fig. 2 Watertight reconstructions under different perturbations. Under small perturbations (first two images from the left), the reconstruction is a triangulation of the sampled manifold. A few non-manifold configurations appear however under larger perturbations (Rightmost image).

acute triangle  $\sigma$ ,  $R_B(\sigma) = R_C(\sigma)$ . Based on [18, 19], we define the total order on 2-simplices:

$$\sigma_{1} \leq \sigma_{2} \iff \begin{cases} R_{B}(\sigma_{1}) < R_{B}(\sigma_{2}) \\ \text{or} \\ R_{B}(\sigma_{1}) = R_{B}(\sigma_{2}) \quad \text{and} \quad R_{C}(\sigma_{1}) \geq R_{C}(\sigma_{2}) \end{cases}$$

$$(4)$$

<sup>399</sup> Under generic condition on the position of points, we can show this order is total. In <sup>400</sup> what follows, the lexicographic order  $\sqsubseteq_{lex}$  is induced by this order on simplices. The <sup>401</sup> following proposition from [19] shows a strong link between the simplex ordering <sup>402</sup> and the 2D Delaunay triangulation.

<sup>403</sup> **Proposition 6** (**Proposition 7.9 in [19**]) Let **P** = {*P*<sub>1</sub>,...,*P<sub>N</sub>*} ⊂ ℝ<sup>2</sup> with *N* ≥ 3 be <sup>404</sup> in general position and let *K*<sub>**P**</sub> be any 2-dimensional complex containing the Delaunay <sup>405</sup> triangulation of **P**. Denote by β<sub>**P**</sub> ∈ *C*<sub>1</sub>(*K*<sub>**P**</sub>) the 1-chain made of edges belonging to <sup>406</sup> the boundary of convex hull CH(**P**). If Γ<sub>min</sub> = min<sub>⊑lex</sub> {Γ ∈ *C*<sub>2</sub>(*K*<sub>**P**</sub>), ∂Γ = β<sub>**P**</sub>}, the <sup>407</sup> simplicial complex |Γ<sub>min</sub>| support of Γ<sub>min</sub> is the Delaunay triangulation of **P**.

As the 2D Delaunay triangulation has some well-known optimality properties, such 408 as maximizing the minimal angle, we can hope that using the same order to minimize 409 2-chains in dimension 3 will keep some of those properties. In fact, it has been shown 410 that for a Čech or Vietoris-Rips complex, under strict conditions linking the point set 411 sampling, the parameter of the complex and the reach of the underlying manifold of 412 Euclidean space, the minimal lexicographic chain using the described simplex order 413 is a triangulation of the sampled manifold [18]. Experimental results (Figure 2) show 414 that this property remains true relatively far from these theoretical conditions. 415

#### 416 6.2 Open surface triangulation

Given a point cloud sampling an open surface and a 1-cycle sampling the boundary of

the surface, we generate a Čech complex of the point cloud using the Phat library [5].

419 The parameter of the complex should be sufficient to capture the homotopy type

<sup>420</sup> of the surface to reconstruct and should contain the provided boundary cycle. After <sup>421</sup> constructing the 2-boundary matrix, we apply the boundary reduction algorithm,

slightly modified to store the transformation matrix V (Section 4.2). We then apply

3

Algorithm 4 to find out if a 2-chain bounded by the cycle exists in the current Čech 423

complex. In this case, we get a chain bounded by the provided cycle and apply 424

Algorithm 3 to minimize the chain under imposed boundary. Otherwise, we might 425 have to increase the Čech parameter to capture the homotopy type of the surface to

- 426 reconstruct [4,11]. Figure 1 shows results of this method.
- 427
- 6.3 Closed surface triangulation 428

Using Algorithm 5 requires a strongly connected 3-pseudomanifold: we therefore use 429

a 3D Delaunay triangulation, for its efficiency and non-parametric nature, using the 430

CGAL library [29], and complete it into a topological 3-sphere by connecting, for any 431

triangle on the convex hull of the Delaunay triangulation, its dual edge to an "infinite" 432

dual vertex. 433

Experimentally, sorting triangles does not require exact predicates: the R<sub>B</sub> and 434

R<sub>C</sub> quantities can simply be calculated in fixed precision. The quasilinear complexity 435

of Algorithm 5 makes it competitive in large point cloud applications. Outliers are 436

naturally ignored and, being parameter free, the algorithm adapts to non uniform point 437

densities, as seen in the closeup of Figure 4. 438

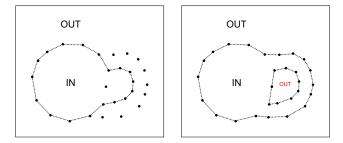


Fig. 3 Providing additional topological information can improve the result of the reconstruction. Here, the lexicographic order on 1-chains is induced by edge length comparison.

The choice of  $\alpha_1$  and  $\alpha_2$  defines the location of the closed separating surface and 439 are chosen interactively. Although we could devise an algorithm where these inputs 440 are not required – the algorithm would simply merge regions until only two connected 441 components remain - this would only work for uniform and non-noisy point clouds 442 but not make for a robust algorithm. On the contrary, adding multiple interior and 443 exterior regions can guide the algorithm by providing better topological constraints, 444 as depicted in Figure 3. 445

Algorithm 5 requires to be slightly modified to take as input multiple  $\alpha_1, \alpha_2$ : after 446 creating all sets with MakeSet, we need to combine all  $\alpha_1$  sets together, and all  $\alpha_2$ 447 sets together. The algorithm remains unchanged for the rest. 448

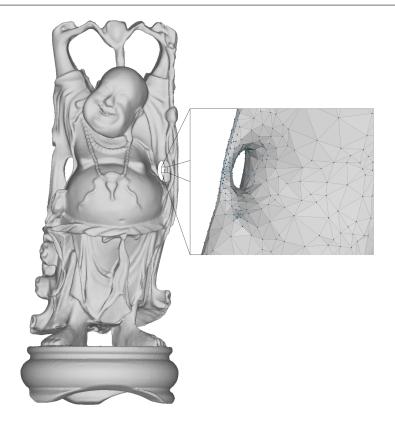


Fig. 4 Closed surface triangulation of 440K points in 7.33 seconds. Beside the point cloud, the only user input is one inner tetrahedron. The closeup shows that small features are correctly recovered.

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