Computational Lab I: Discrete Algebraic Surface Topology

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The discrete version of algebraic surface topology was built on oriented 2-dimensional simplicial complex, namely the triangular mesh, which was implemented by half-edge data structure in computer.

1 Half-edge Data Structure

Definition 1 (Half-edge Data Structure). The *half-edge* data structure of triangular mesh approximating an oriented surface has the following classes:

V *vertex* class

H *half-edge* class, oriented from one vertex (the *source* vertex) to another vertex (the *target* vertex)

E *edge* class, each edge has two opposite half-edges, with the exception that the edge on boundary only has one half-edge

F *face* class, each face has three half-edges, oriented counter-clockwisely with respect to the normal of face

with the following pointer functions which take in the realization of classes above:

$v(\cdot)$ vertex pointer function $v : \mathbf{H} \to \mathbf{V}$ parametrized by "source/target" that points a half-edge to a vertex. $v_{\text{sour}}(\cdot)$ points to the source vertex, and $v_{\text{targ}}(\cdot)$ points to the target vertex

$f(\cdot)$ face pointer function $f : \mathbf{H} \to \mathbf{F}$ that points a half-edge to a face that it belongs to

$e(\cdot)$ edge pointer function $e : \mathbf{H} \to \mathbf{E}$ that points a half-edge to an edge that it belongs to

$h(\cdot)$ the polymorphic half-edge pointer function can recognize different inputs and take different actions:

half-edge $h : \mathbf{H} \to \mathbf{H}$ parametrized by "next/previous" that points a half-edge to another half-edge, where $h_{\text{next}}(\cdot)$ points to the next half-edge, and $h_{\text{prev}}(\cdot)$ points to the previous half-edge

face $h : \mathbf{F} \to \mathbf{H}$ points a face to the *first* half-edge it contains

vertex $h : \mathbf{V} \to \mathbf{H}$ points a vertex to the *first* half-edge that the vertex was targeted

edge $h : \mathbf{E} \to \mathbf{H} \times \mathbf{H}$ points an edge to the pair of half-edges it contains (with exception of boundary edge that has only one half-edge, in that case $h : \mathbf{E} \to \mathbf{H}$)

Example 2 (2-chain). As shown in figure 1, we have realization of classes:

$$\mathbf{V} = \{v_1, v_2, v_3, v_4, v_5, v_6\} =: v_{1:6}$$
$$\mathbf{H} = h_{1:12}$$
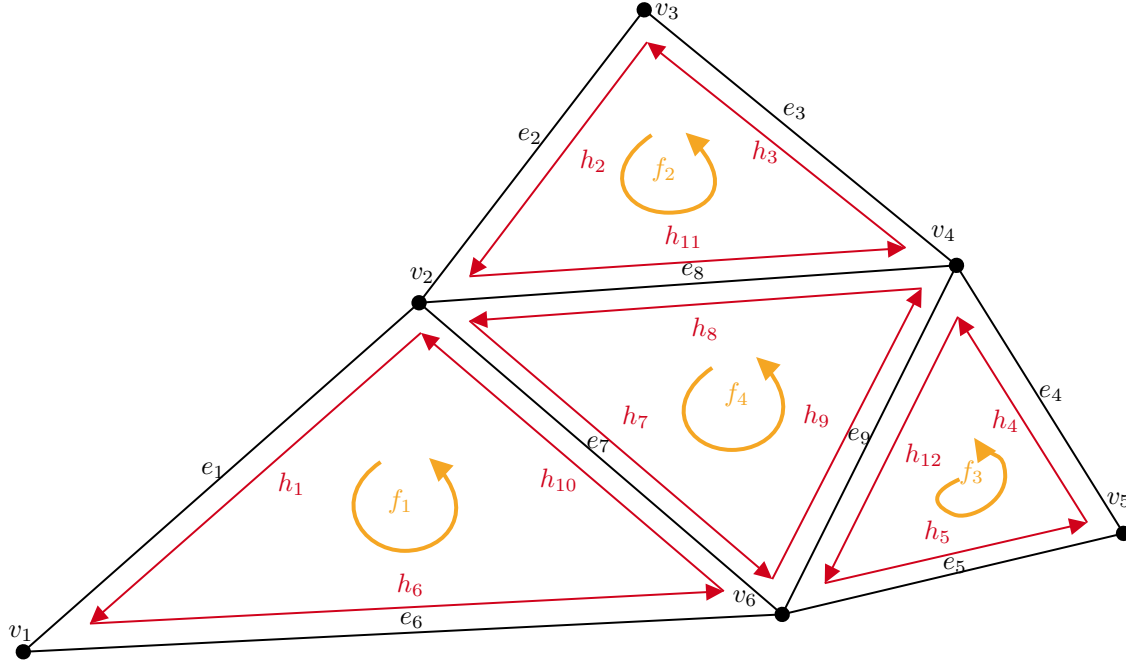$$\mathbf{E} = e_{1:9}$$
$$\mathbf{F} = f_{1:4}$$

Figure 1: an example of half-edge data structure of a simplicial complex $\boldsymbol{\Sigma} = f_1 + f_2 + f_3 + f_4$

and pointer functions :

1. $v(\cdot)$ is obvious to every half-edge in the picture, e.g. $v_{\mathrm{sour}}(h_1) = v_2$ and $v_{\mathrm{targ}}(h_1) = v_1$

2. $f(\cdot)$ is also obvious to every face in the picture, e.g. $f(h_1) = f(h_6) = f(h_{10}) = f_1$

3. $e(\cdot)$ in this picture is:

$$e(h_n) = \begin{cases} e_n & \text{if } n \leq 9 \\ e_{n-3} & \text{otherwise} \end{cases}$$

4. $h(\cdot)$ in this picture, firstly

$$h(e_n) = \begin{cases} h_n & \text{if } e_n \text{ on boundary} \\ (h_n, h_{n+3}) & \text{otherwise} \end{cases}$$

secondly we denote e.g. $h_{\mathrm{next}}(h_1) = h_6$ and $h_{\mathrm{prev}}(h_6) = h_1$ as $1 \to 6$, then we denote all $h : \mathbf{H} \to \mathbf{H}$ as

$$1 \to 6 \to 10 \to 1 \qquad 2 \to 11 \to 3 \to 2 \qquad 4 \to 12 \to 5 \to 4 \qquad 8 \to 7 \to 9 \to 8$$

thirdly we set "the first half-edge" of each face and fourthly "the first half-edge" of each target vertex in this picture (note that these are merely arbitrary choices) as follows:

$$h(f_1) = h_1 \quad h(f_2) = h_2 \quad h(f_3) = h_5 \quad h(f_4) = h_8$$

$$h(v_n) = h_n$$

Now we rephrase example 2 in discrete surface topology language.

2   Discrete Algebraic Surface Topology

Definition 3 (Simplex). Suppose $k + 1$ linear independent points embedded in $\mathbb{R}^n$

$$v_0, v_1, ..., v_k$$

the standard *k-simplex*

$$[v_0, v_1, ..., v_k]$$

is the minimal convex set including all of them.

e.g. in figure 2, the 2-simplex (vertices written counter-clock-wisely)

$$f_1 = [v_2, v_1, v_6]$$

the 1-simplex (vertices written from source to target)

$$h_1 = [v_2, v_1]$$

Definition 4 (Simplicial Complex). A *simplicial complex* $\boldsymbol{\Sigma}$ is an union of simplicies with "vertex auto -alignment".

Definition 5 (Chain). A *k-chain* is **a linear combination (formal sum)** of all k-simplicies in $\boldsymbol{\Sigma}$.

Definition 6 (Chain Space). The *k-dimensional chain space* is the linear space formed by all k-chain by formal sum over $\mathbb{Z}$, denoted as

$$C_k(\boldsymbol{\Sigma}, \mathbb{Z})$$

e.g. in figure 2,

- in $C_2(\boldsymbol{\Sigma}, \mathbb{Z})$, a 2-chain could be f_1, $f_1 + f_2$, or even $f_1 + 3f_2 - 4f_3$

- in $C_1(\boldsymbol{\Sigma}, \mathbb{Z})$, a 1-chain could be $h_1 + h_2$, or even $2h_1 - 5h_2 + 3h_4$

Definition 7 (Boundary Operator). The discrete version of boundary operator ∂, could be implemented as follows: we use hat notation as we delete $\hat{v}_i$ from $[v_0, ..., \hat{v}_i, ..., v_k]$. Then

$$\partial[v_0, v_1, ..., v_k] = \sum_{i=0}^{k} (-1)^i [v_0, ..., \hat{v}_i, ..., v_k]$$

e.g. in figure 2,

- $\partial f_2 = \partial[v_3, v_2, v_4] = [v_3, v2] + [v_2, v_4] - [v_3, v_4] = h_2 + h_{11} + h_3$

- $\partial h_1 = \partial[v_2, v_1] = v_1 - v_2$

we would easily see that the boundary operator is linear, e.g.

$$\partial \boldsymbol{\Sigma} = \partial(f_1 + f_2 + f_3 + f_4) = \partial f_1 + \partial f_2 + \partial f_3 + \partial f_4$$

and $\partial^2 = \mathbf{0}$, e.g.

$$\partial^2 f_2 = \partial(h_2 + h_{11} + h_3) = v_2 - v_3 + v_4 - v_2 + v_3 - v_4 = \mathbf{0}$$

Definition 8 (Cochain Space). By assigning a value to each k-simplex (opposite half-edges get same value with different symbols), one can get a *k-dimensional cochain space* formed by all k-cochain, which takes in a k-chain and outputs the summation of value of k-simplicies of the k-chain.

We define discrete version:

$$\text{closed 1-chain} \longleftrightarrow \text{loop}$$
$$\text{exact 1-chain} \longleftrightarrow \text{boundary}$$
$$\text{closed 1-cochain} \longleftrightarrow \text{curl free field}$$
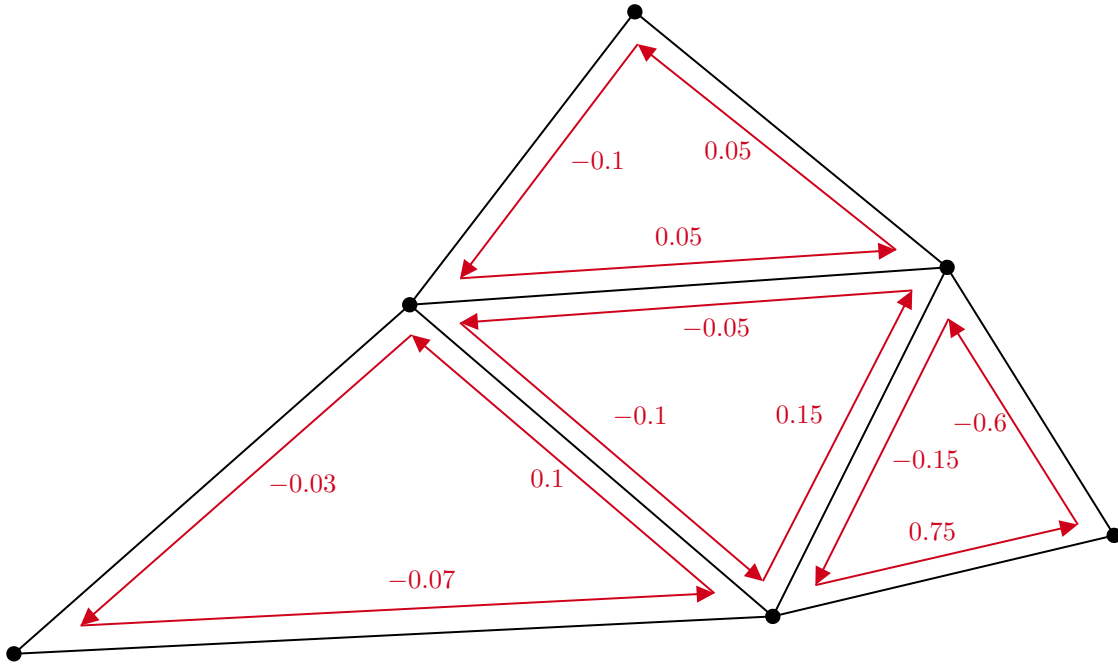$$\text{exact 1-cochain} \longleftrightarrow \text{gradient field}$$

Figure 2: an example of 1-cochain w constructed from Σ

Example 9 (1-cochain). see figure 2, we have

$$w(h_2 + h_7) = w(h_2) + w(h_7) = -0.1 - 0.1 = -0.2$$

and one can see that any exact 1-chain γ in Σ, we have

$$w(\gamma) = 0$$

so w is an exact 1-cochain.

Recall that *graph* $G = (V, E)$, by its definition, is contained in a simplicial complex Σ.

Definition 10 (Dual Graph)**.** The *dual graph* $\bar{G}$ of a plane graph G, is a graph that

- has a vertex for each face of G

- has an edge whenever two faces of G are separated from each other by an edge, and a self-loop when the same face appears on both sides of an edge

Thus, each edge e of G has a corresponding dual edge $\bar{e}$ in $\bar{G}$, whose endpoints are the dual vertices corresponding to the faces on either side of e.

Definition 11 (Spanning Tree)**.** A *spanning tree* T of an undirected graph G is a subgraph that is a tree which includes all of the vertices of G, with a minimum possible number of edges.

3 Algorithms of $\pi_1(\Sigma)$, $\tilde{\Sigma}$, $H_1(\Sigma)$ and $H^1(\Sigma)$

We now introduce algorithms to compute

- $\pi_1(\Sigma)$, first homotopy group of Σ

- $\tilde{\Sigma}$, universal covering space Σ

- $H_1(\Sigma)$, first homology group of Σ

- $H^1(\Sigma)$, first cohomology group of Σ

represented by 1-chain or 1-cochain as group basis. For computation of the first homology group $H_1(\Sigma)$, it has the same bases with the first homotopy group $\pi_1(\Sigma)$. However, it is not the case in higher dimensions. For higher dimensional computation of homology group basis, please consult eigen-decomposition of combinatorial Laplace operator using Smith norm. One would be able to finish assignment 1 from David Gu[1] now.
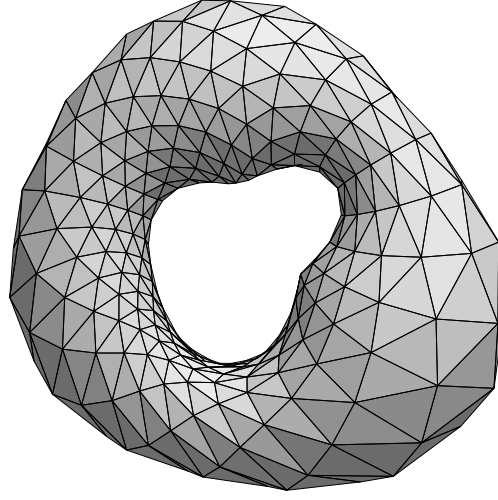
4 Algorithm of Obstruction Class

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[1] https://www3.cs.stonybrook.edu/~gu/lectures/2020/

Figure 3: a triangulation mesh on a genus 1 surface

Algorithm 1: First Homotopy Group $\pi_1(\mathbf{\Sigma})$

Input: a closed triangular mesh $\mathbf{\Sigma}$
Output: $\pi_1(\mathbf{\Sigma})$

1. compute the dual mesh $\bar{\mathbf{\Sigma}}$ of the input mesh $\mathbf{\Sigma}$

2. compute a spanning tree $\bar{\mathbf{T}}$ of $\bar{\mathbf{\Sigma}}$, rooted at an arbitrary point p

3. the *cut graph* $\mathbf{\Gamma}$ of $\mathbf{\Sigma}$ is given by
$$\mathbf{\Gamma} := \{e \in \mathbf{\Sigma} \mid \bar{e} \notin \bar{\mathbf{T}}\}$$

4. compute a spanning tree $\mathbf{T}$ of $\mathbf{\Gamma}$

5. select an edge $e_i \in \mathbf{\Gamma} \setminus \mathbf{T}$, then $e_i \cup \mathbf{T}$ gives an unique closed 1-chain γ_i; suppose we get a set of distinct 1-chains
$$\{\gamma_1, \gamma_2, ..., \gamma_k\}$$
which is the set of generators of $\pi_1(\mathbf{\Sigma})$

6. cut the mesh $\mathbf{\Sigma}$ along $\mathbf{\Gamma}$ to obtain $\tilde{\mathbf{\Sigma}}_0$, the fundamental domain of $\mathbf{\Sigma}$

7. set $R = \emptyset$, let $\gamma = \partial\tilde{\mathbf{\Sigma}}_0$, traverse γ, once $e_i^{\pm 1}$ is encountered, append $\gamma_i^{\pm 1}$ to R,
$$R \leftarrow R\gamma_i^{\pm 1}$$

8. the first homotopy group of $\mathbf{\Sigma}$
$$\pi_1(\mathbf{\Sigma}, p) = \langle \gamma_1, \gamma_2, ..., \gamma_k \mid R \rangle$$

Algorithm 2: Universal Covering Space $\tilde{\boldsymbol{\Sigma}}$

Input: a closed triangular mesh $\boldsymbol{\Sigma}$

Output: an universal covering space $\tilde{\boldsymbol{\Sigma}}$ with desired size

1. the same as step $1 \to 6$ in algorithm 1

2. set $\tilde{\boldsymbol{\Sigma}} = \tilde{\boldsymbol{\Sigma}}_0$, glue a copy of $\tilde{\boldsymbol{\Sigma}}_0$ with $\tilde{\boldsymbol{\Sigma}}$ along γ_i, a homeomorphism $h : \partial\tilde{\boldsymbol{\Sigma}} \supset \gamma_i \sim \gamma_i^{-1} \subset \partial\tilde{\boldsymbol{\Sigma}}_0$

$$\tilde{\boldsymbol{\Sigma}} \leftarrow \tilde{\boldsymbol{\Sigma}} \cup_h \tilde{\boldsymbol{\Sigma}}_0$$

3. trace the boundary of $\tilde{\boldsymbol{\Sigma}}$, if there are two adjacent 1-chains $\gamma_j, \gamma_{j+1} \subset \partial\tilde{\boldsymbol{\Sigma}}$, such that $\gamma_j^{-1} = \gamma_{j+1}$ then glue them together

4. repeat step 2 and step 3, until $\tilde{\boldsymbol{\Sigma}}$ is large enough

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Algorithm 3: First Homology Group $H_1(\boldsymbol{\Sigma})$

Input: a closed triangular mesh $\boldsymbol{\Sigma}$

Output: $H_1(\boldsymbol{\Sigma})$

1. the same as step $1 \to 5$ in algorithm 1

2. $H_1(\boldsymbol{\Sigma}) = \{[\gamma_1], [\gamma_2], ..., [\gamma_{2g}]\}$

Algorithm 4: First Cohomology Group $H^1(\boldsymbol{\Sigma})$

Input: a closed triangular mesh $\boldsymbol{\Sigma}$

Output: $H^1(\boldsymbol{\Sigma})$

1. the same as step $1 \to 2$ in algorithm 3

2. for each γ_i, slice $\boldsymbol{\Sigma}$ along γ_i to obtain a mesh $\boldsymbol{\Sigma}_i$ with two boundaries. We have $\partial\boldsymbol{\Sigma}_i = \gamma_i^+ - \gamma_i^-$

3. set a 0-form τ_i on $\boldsymbol{\Sigma}_i$ such that $\tau_i(v^+) = 1$ for all vertices $v^+ \in \gamma_i^+$ and $\tau_i(v^-) = 0$ for all vertices $v^- \in \gamma_i^-$; set $w_i = d\tau_i$

4. $H^1(\boldsymbol{\Sigma}) = \{[w_1], [w_2], ..., [w_{2g}]\}$