

# Analysis of the inference of ratings and rankings on Higher Order Networks with complex topologies

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Bla bla...

## I. INTRODUCTION

See [1, 2].

## II. THEORY

### A. Least-squares inference of team ratings and rankings

Consider a set of  $N = n + 1$  teams labeled  $\{0, 1, \dots, n\}$ . For each distinct pair of teams  $i$  and  $j$ , assume a pairwise antisymmetric comparison score  $f_{ij} = -f_{ji}$  is available. Informally,  $f_{ij} > 0$  indicates that team  $i$  is rated higher than team  $j$ , whereas  $f_{ij} < 0$  suggests the opposite.

A natural approach to inferring consistent ratings is to assign scores  $w_i$  ( $i = 0, 1, \dots, n$ ) that minimize the total squared discrepancy

$$\sum_{i < j} (f_{ij} - (w_i - w_j))^2. \quad (1)$$

Once the team ratings  $w_0, w_1, \dots, w_n$  have been obtained, the ranking  $r_0, r_1, \dots, r_n$  follows from the ordering condition  $r_i < r_j$  whenever  $w_i > w_j$ . This least-squares formulation defines a well-posed optimization problem that admits an elegant reformulation in terms of cochains over simplicial complexes. The resulting framework for rating and ranking inference is known as *HodgeRank* [1, 3].

### B. Discrete exterior calculus on simplicial complexes

To review the HodgeRank method, we recall the basic elements of discrete exterior calculus on simplicial complexes. Let  $V = \{0, 1, \dots, n\}$  denote a set of vertices, and define a simplicial complex as a collection

$$K = K_0 \cup K_1 \cup \dots \cup K_n,$$

where

$$K_k \subseteq \{\{i_0, \dots, i_k\} : 0 \leq i_0 < \dots < i_k \leq n\}$$

is the set of  $k$ -simplices, i.e., unordered subsets  $\{i_0, \dots, i_k\}$  of vertices satisfying the closure condition: if  $s \in K$  and  $r \subset s$  with  $r \neq \emptyset$ , then  $r \in K$ .

The set  $K$  of simplices is associated with a vector space  $C$  of *chains*, with canonical basis  $\{e_s : s \in K\}$ . We equip  $C$  with the inner product

$$e_r \cdot e_s = \delta_{rs}, \quad r, s \in K.$$

With Einstein summation convention, any chain  $c \in C$  can be written as  $c = c^s e_s$  with coefficients  $c^s \in \mathbb{R}$ . The chain space decomposes as

$$C = C_0 \oplus C_1 \oplus C_2,$$

where  $C_k$  is spanned by  $\{e_s : s \in K_k\}$ .

The dual space  $C^*$  of *cochains* is spanned by  $\{e^s : s \in K\}$ , where  $e^r(e_s) = \delta^r_s$ . Any  $f \in C^*$  decomposes as  $f = f_s e^s$  with coefficients  $f_s \in \mathbb{R}$ , and

$$C^* = C_0^* \oplus C_1^* \oplus C_2^*.$$

By Riesz's theorem, every cochain  $f \in C^*$  corresponds uniquely to a chain  $f^\flat \in C$  such that  $f(c) = f^\flat \cdot c$  for all  $c \in C$ . This induces an inner product on  $C^*$ :

$$f \cdot g = f^\flat \cdot g^\flat.$$

The map  $f \mapsto f^\flat$  is an isomorphism, with inverse denoted  $c \mapsto c^\sharp$ .

**Orientation and boundary.** An orientation is fixed by requiring

$$e_{i_0 \dots i_k} = (-1)^p e_{j_0 \dots j_k},$$

for each simplex  $s = \{i_0, \dots, i_k\} \in K$ , where  $p$  is the parity of the permutation mapping  $(i_0, \dots, i_k)$  to  $(j_0, \dots, j_k)$ .

The boundary operator  $\partial : C \rightarrow C$  is defined on basis elements as

$$\partial e_{i_0 \dots i_k} = \sum_{j=0}^k (-1)^j e_{i_0 \dots i_{j-1} i_{j+1} \dots i_k}.$$

Thus  $\partial(C_k) \subseteq C_{k-1}$ , with  $C_{-1} = \{0\}$ , yielding the decomposition

$$\partial = \partial_0 \oplus \partial_1 \oplus \partial_2, \quad \partial_k(C_k) \subseteq C_{k-1}, \quad \partial_0 = 0.$$

**Coboundary.** The coboundary operator  $d : C^* \rightarrow C^*$  is defined by

$$df(c) = f(\partial c), \quad f \in C^*, \quad c \in C.$$

Hence  $d(C_k^*) \subseteq C_{k+1}^*$ , with  $C_3^* = \{0\}$ , and

$$d = d_0 \oplus d_1 \oplus d_2, \quad d_2 = 0.$$

**Adjoint operators.** The adjoints  $\partial^* : C \rightarrow C$  and  $d^* : C^* \rightarrow C^*$  are defined by

$$\partial^* c \cdot q = c \cdot \partial q, \quad d^* f \cdot g = f \cdot dg,$$

for all  $c, q \in C$  and  $f, g \in C^*$ . They decompose as

$$\partial^* = \partial_0^* \oplus \partial_1^* \oplus \partial_2^*, \quad d^* = d_0^* \oplus d_1^* \oplus d_2^*.$$

**Hodge Laplacians.** The boundary and coboundary operators define the Hodge Laplacians

$$\begin{aligned} \mathcal{L} &= (\partial + \partial^*)^2 = \partial\partial^* + \partial^*\partial : C \rightarrow C, \text{ and} \\ L &= (d + d^*)^2 = dd^* + d^*d : C^* \rightarrow C^*, \end{aligned}$$

with  $\partial^2 = (\partial^*)^2 = d^2 = (d^*)^2 = 0$ . These decompose as

$$\mathcal{L} = \mathcal{L}_0 \oplus \mathcal{L}_1 \oplus \mathcal{L}_2, \quad L = L_0 \oplus L_1 \oplus L_2,$$

with

$$\mathcal{L}_k = \mathcal{L}_k^\downarrow \oplus \mathcal{L}_k^\uparrow, \quad L_k = L_k^\downarrow \oplus L_k^\uparrow,$$

$$\mathcal{L}_k^\downarrow = \partial_k^* \partial_k, \quad \mathcal{L}_k^\uparrow = \partial_{k+1} \partial_{k+1}^*,$$

$$L_k^\uparrow = d_k^* d_k, \quad L_k^\downarrow = d_{k-1} d_{k-1}^*.$$

**Hodge decomposition.** Every  $k$ -cochain  $f$  admits the Hodge decomposition

$$f = g \oplus h \oplus s,$$

where  $g \in \text{im } d_{k-1}$  (gradient),  $h \in \ker L_k$  (harmonic), and  $s \in \text{im } d_k^*$  (solenoidal). Equivalently, there exist  $w \in C_{k-1}^*$  and  $u \in C_{k+1}^*$  such that

$$g = d_{k-1} w, \quad s = d_k^* u,$$

with  $w$  and  $u$  determined by

$$d_k f = L_{k+1}^\downarrow u, \quad (2)$$

$$d_{k-1}^* f = L_{k-1}^\uparrow w, \quad (3)$$

while  $h = f - g - s$ .

The least-squares problem (1) admits a reformulation in the language of discrete calculus:

$$\|f - d_0(w)\|, \quad (4)$$

where the scalar entries  $f_{ij}$  of the 1-cochain  $f = f_{ij} e^{ij}$  encode the pairwise comparison scores, and the components  $w_i$  of the 0-cochain  $w = w_i e^i$  represent the team ratings. The minimizer of (4) coincides with the solution of Eq. (3) for  $k = 1$ .

In this setting, the set of 0-simplices  $K_0$  corresponds to the teams  $V$ , the set of 1-simplices  $K_1$  to the pairwise comparisons  $E$ , and the set of 2-simplices  $K_2$  to the closed triples of comparisons:

$$K_2 = \{\{i, j, k\} : \{i, j\}, \{j, k\}, \{i, k\} \in E\}.$$

For  $k > 2$ ,  $K_k = \emptyset$ .

### III. METHODS

The dataset covers five major soccer leagues: England, France, Germany, Italy, and Spain. Each league consists of  $N = 20$  teams, except Germany, which has  $N = 18$ .

In these competitions (e.g., Premier League, La Liga, Serie A, Bundesliga, Ligue 1), the final standings are determined by a points system based on match results throughout the season. In the standard system, a win yields 3 points, a draw 1 point, and a loss 0 points. Each team plays every other team twice (home and away) in a round-robin format. Over the season, the  $i$ -th team accumulates  $\pi_i$  total points. At the end of the season, teams are ranked in descending order of  $\pi_i$ . If two or more teams tie on points, tiebreakers are applied, typically in the following order (with minor variations across leagues): goal difference (GD: goals scored minus goals conceded), goals scored (GF), head-to-head record, and, in rare cases, a playoff or extra match. Thus, the final table is determined primarily by total points, with tiebreakers resolving ties.

For each pair  $\{i, j\}$  of teams in a given league, several pairwise antisymmetric comparison scores  $f_{ij}$  are defined:

1. **crossings:** ...
2. **counterattacks:** ...
3. **pressure loss:** ...
4. **build-up time:** ...
5. **direct play:** ...
6. **pressure points:** ...
7. **shots:** ...
8. **flow rate:** ...
9. **maintenance time:** ...
10. **middle-zone time:** ...

11. **style:** the Singular Value Decomposition (SVD) of previous metrics is used to define higher-level descriptors called *styles*, so  $f_{ij} = \dots$  [2].
12. **rank based ground-truth:**  $f_{ij} = i - j$ .
13. **point based ground-truth:**  $f_{ij} = \pi_j - \pi_i$ .

#### A. HodgeRank inference via metrics

Given a metric, the coefficients

$$f_{ij} = -\frac{1}{2}(m_{ij} + m_{ji})$$

are defined for each  $i < j$ . The minus sign is imposed to identify ratings as potentials [1]. To test the HodgeRank inference method, two types of ground-truth coefficients are also introduced: (i) *rank-based* coefficients  $f_{ij} = i - j$ , and (ii) *points-based* coefficients  $f_{ij} = \pi_j - \pi_i$ .

In all cases, the underlying structure is a fully connected graph with  $N$  nodes, promoted to a simplicial complex composed of 0-simplices labeled by  $i$ , 1-simplices labeled by  $ij$  with  $i < j$ , and 2-simplices labeled by  $ijk$  with  $i < j < k$ . Correspondingly, the coefficients  $f_{ij}$  form the components of a 1-cochain  $f$ , where the anti-symmetry  $f_{ji} = -f_{ij}$  is imposed by definition. There are no  $k$ -simplices for  $k > 2$ .

#### IV. RESULTS

The following experiments are carried on.

1. Compute ratings and rankings from the full weighted network of scores  $\phi_{ij}$ .
2. Compute ratings and rankings from sparsified version of the weighted networks of scores  $\phi_{ij}$ , by randomly removing a varying fraction of links.
3. Compute ratings and rankings from sparsified version of the weighted networks, by removing a varying fraction of the weighted links of the weighted matrix of scores  $\phi_{ij}$  sorted in decreasing order of absolute weight.
4. Compute ratings defining the 1-cochain coordinates  $f'_{ij} = \phi_j - \phi_i$ , where the strength  $\phi_i = \sum_j w_{ij}$  defines a local potential for the  $i$ -th team.

#### V. CONCLUSIONS

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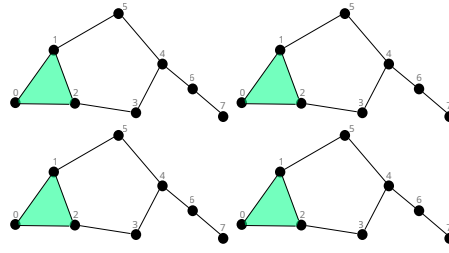


FIG. 1. (color online).