## Hypergraph Matching

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## 1 Problem Setup

Consider two correlated Erdős-Rényi m-hypergraphs  $\mathcal{A}, \mathcal{B} \sim \mathcal{G}(n, q)$  on the same vertex sets [n]. Let  $\pi : [n] \mapsto [n]$  denote the latent permutation. Assume that conditional on  $\mathcal{A}$ , for all  $\omega = (\omega_1, \ldots, \omega_m)$  where  $1 \leq \omega_1 < \cdots < \omega_m \leq n$ ,  $\mathcal{B}_{\pi(\omega)}$  are independent and distributed as

$$B_{\pi(\omega)} \sim \begin{cases} Ber(s) & \text{if} \quad \mathcal{A}_{\omega} = 1\\ Ber\left(\frac{q(1-s)}{1-q}\right) & \text{if} \quad \mathcal{A}_{\omega} = 0. \end{cases}$$

Note that  $\mathcal{A}, \mathcal{B}$  are symmetric tensors, and  $\mathcal{A}(\omega) = \mathcal{B}(\omega) = 0$  for all the entries  $\omega$  in which  $\omega_i = \omega_j$  for some  $i \neq j \in [n]$ .

## 2 Derivation of distance statistics

Define the open neighbourhood of i as the collection of other nodes that are connected to node i

$$\mathcal{N}_{\mathcal{A}}(i) = \{(i_2, \dots, i_m) \in [n]^{m-1} : \mathcal{A}_{i, i_2, \dots, i_m} = 1\}.$$

Define the plain neighbourhood of i as

$$\mathcal{C}_{\mathcal{A}}(i) = \{i \in [n] : i \in \omega, \omega \in \mathcal{N}_{\mathcal{A}}(i)\} \cup \{i\},\$$

where  $i \in \omega$  is equivalent to  $i = \omega_j$  for some  $j \in [n]$ . Then  $\mathcal{C}_{\mathcal{A}}(i)$  collects all the nodes in [n] connected with i via a higher-order connection  $\mathcal{A}_{\omega}$  for some  $\omega \in \mathcal{N}_{\mathcal{A}}(i)$ . Let  $c_a(i) = |\mathcal{C}_{\mathcal{A}}(i)|$ .

Define the "innerneighbourhood" of vertex j of i as

$$\mathcal{D}_{\mathcal{A}}(j)^{(i)} = \{(i_1, \dots, i_{m-1}) : i_k \neq i_l, \text{ for all } k \neq l \in [m-1], i_k \in \mathcal{C}_{\mathcal{A}}(i)/\{j\} \text{ for all } k \in [m-1]\}$$
$$\cup \{(i_1, \dots, i_{m-1}) : \text{at least one of } i_k = j\}.$$

Note that  $\mathcal{D}_{\mathcal{A}}(j)^{(i)}$  collects all the sub-edges that (1) are generated from the plain neighbourhood of i and (2) leads to non-defined hyperedges involving at least two elements equal to j, e.g.,  $(j, 1, \ldots, j)$ . Also note that

$$|\mathcal{D}_{\mathcal{A}}(j)^{(i)}| = P_{c_a(i)}^{m-1} + \sum_{k=1}^{m-1} {m-1 \choose k} n^{m-1-k},$$

where  $P_{c_a(i)}^{m-1}$  is the number of permutation of m-1 elements out of  $c_a(i)$  elements.

Then, the "outdegree" of vertex j is

$$a_{j}^{(i)} = \frac{1}{\sqrt{n^{m-1} - |\mathcal{D}_{\mathcal{A}}(j)^{(i)}|}} \sum_{\substack{\omega \notin \mathcal{D}_{\mathcal{A}}(j)^{(i)} \\ \omega \notin \mathcal{D}_{\mathcal{A}}(j)^{(i)}}} (\mathcal{A}_{\omega,j} - q)$$

$$= \frac{1}{\sqrt{n^{m-1} - (P_{c_{a}(i)}^{m-1} + \sum_{k=1}^{m-1} {m-1 \choose k} n^{m-1-k})}} \sum_{\substack{\omega \notin \mathcal{D}_{\mathcal{A}}(j)^{(i)} \\ \omega \notin \mathcal{D}_{\mathcal{A}}(j)^{(i)}}} (\mathcal{A}_{\omega,j} - q).$$

Next, note that we need to consider all the  $j \in \mathcal{C}_{\mathcal{A}}(i)$  except the node i itself. So, we define the empirical distribution

$$\bar{\mu}_i = \frac{1}{c_a(i) - 1} \sum_{\mathcal{C}_{\mathcal{A}}(i)/\{i\}} \delta_{a_j^{(i)}} - \overline{\text{Binom}}(n^{m-1} - |\mathcal{D}_{\mathcal{A}}(j)^{(i)}|, q).$$

Similar for network  $\mathcal{B}$ .

## References