

Notes on a generative model of hypergraphs

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1 Problem

We want to find a way of generating hypergraphs with prescribed expected vertex degrees. We proceed to find a generalization of the, so-called, Chung-Lu model for hypergraphs. This will be done using an exponential random hypergraph formulation. We begin by considering 3-regular hypergraphs. The result is then generalized to g -regular hypergraphs.

A new part that was added in August, extends the results to g -regular hypergraphs with α -dimensional joint prescribed expected vertex degrees.

On September 12th a small correction was made that added a factor of $\frac{1}{2}$ to Eqs. 83 through 84. and Eqs. 92 through 95. This change did not affect any other part. This is version 3 of these notes.

2 Prescribed Expected Vertex Degrees

2.1 3-regular Hypergraphs

We wish to randomly generate hypergraphs in with prescribed expected degrees

$$\bar{\mathbf{k}} = \{\bar{k}_1, \bar{k}_2, \dots, \bar{k}_N\} \quad (1)$$

where there are assumed to be N vertices. Calling the ensemble of all possible simple 3-regular hypergraphs \mathcal{G}_3 , the expected degree of vertex i is

$$\bar{k}_i = \sum_{G \in \mathcal{G}_3} k_i(G) P(G) \quad (2)$$

where $k_i(G)$ is the degree of vertex i in a graph G and $P(G)$ is the probability of generating that graph.

Maximizing entropy while attempting to enforce the prescribed degree constraints, the probability of generating a hypergraph is

$$P(G; \vec{\beta}) = \frac{1}{Z(\vec{\beta})} e^{-\vec{\beta} \cdot \mathbf{k}} \quad (3)$$

where $\vec{\beta}$ is a vector of Laplace coefficients, and \mathbf{k} is the vector of vertex degrees. In what follows we will choose the elements of $\vec{\beta}$ so that the prescribed degree constraints are enforced.

Now, a 3-regular hypergraph can be represented by a 3-dimensional adjacency tensor \mathbf{a} , with elements of

$$a_{ijk} = \begin{cases} 1 & \text{If a hyperedge exists connecting vertices } i, j, \text{ and } k. \\ 0 & \text{Otherwise.} \end{cases} \quad (4)$$

Require, of course, that

$$a_{ijk} = a_{ikj} = a_{jik} = a_{jki} = a_{kij} = a_{kji}, \quad \forall i, j, k, \quad (5)$$

and that

$$a_{ijj} = a_{jij} = a_{jji} = 0, \quad \forall i, j. \quad (6)$$

Note that

$$k_i = \sum_{j=1}^N \sum_{k=1}^N a_{ijk} \quad (7)$$

Recognizing Eqn. 3 as a Gibbs distribution, identify the “Hamiltonian”

$$\mathcal{H}(G; \vec{\beta}) = \vec{\beta} \cdot \mathbf{k} \quad (8)$$

$$= \sum_{i=1}^N \beta_i k_i \quad (9)$$

$$= \sum_{i=1}^N \beta_i \sum_{j=1}^N \sum_{k=1}^N a_{ijk} \quad (10)$$

$$= \sum_{i=1}^N \sum_{j=1}^N \sum_{k=1}^N \beta_i a_{ijk} \quad (11)$$

$$= \sum_{i < j < k} \beta_i a_{ijk} + \sum_{i < k < j} \beta_i a_{ijk} + \sum_{j < i < k} \beta_i a_{ijk} \\ + \sum_{j < k < i} \beta_i a_{ijk} + \sum_{k < i < j} \beta_i a_{ijk} + \sum_{k < j < i} \beta_i a_{ijk} \quad (12)$$

$$= 2 \left(\sum_{i < j < k} \beta_i a_{ijk} + \sum_{i < j < k} \beta_j a_{ijk} + \sum_{i < j < k} \beta_k a_{ijk} \right) \quad (13)$$

$$= 2 \sum_{i < j < k} (\beta_i + \beta_j + \beta_k) a_{ijk} \quad (14)$$

The “partition function” then is

$$\mathcal{Z}(\vec{\beta}) = \sum_{G \in \mathcal{G}_3} e^{-\mathcal{H}(G; \vec{\beta})} \quad (15)$$

$$= \sum_{a_{123}=0}^1 \sum_{a_{124}=0}^1 \cdots \sum_{a_{N-2, N-1, N}=0}^1 e^{-2 \sum_{i < j < k} (\beta_i + \beta_j + \beta_k) a_{ijk}} \quad (16)$$

$$= \prod_{i < j < k} \sum_{a_{ijk}=0}^1 e^{-2(\beta_i + \beta_j + \beta_k) a_{ijk}} \quad (17)$$

$$= \prod_{i < j < k} \left(1 + e^{-2(\beta_i + \beta_j + \beta_k)} \right) \quad (18)$$

The probability that a hyperedge exists between vertices r , s , and t is (assume that $r < s < t$)

$$p_{rst} = \langle a_{rst} \rangle \quad (19)$$

$$= \sum_{G \in \mathcal{G}_3} a_{rst}(G) P(G; \vec{\beta}) \quad (20)$$

$$= \frac{1}{\mathcal{Z}(\vec{\beta})} \sum_{a_{123}=0}^1 \sum_{a_{124}=0}^1 \cdots \sum_{a_{N-2, N-1, N}=0}^1 a_{rst} e^{-2 \sum_{i < j < k} (\beta_i + \beta_j + \beta_k) a_{ijk}} \quad (21)$$

$$= \frac{\left[\prod'_{i < j < k} \left(1 + e^{-2(\beta_i + \beta_j + \beta_k)} \right) \right] \sum_{a_{rst}=0}^1 a_{rst} e^{-2(\beta_r + \beta_s + \beta_t) a_{rst}}}{\prod_{i < j < k} \left(1 + e^{-2(\beta_i + \beta_j + \beta_k)} \right)} \quad (22)$$

where the prime on the product in the numerator indicates a product over all vertex combinations *except* rst . Continuing,

$$p_{rst} = \frac{\sum_{a_{rst}=0}^1 a_{rst} e^{-2(\beta_r + \beta_s + \beta_t) a_{rst}}}{1 + e^{-2(\beta_r + \beta_s + \beta_t)}} \quad (23)$$

$$= \frac{e^{-2(\beta_r + \beta_s + \beta_t)}}{1 + e^{-2(\beta_r + \beta_s + \beta_t)}} \quad (24)$$

$$= \frac{1}{1 + e^{2(\beta_r + \beta_s + \beta_t)}} \quad (25)$$

In the sparse hypergraph limit require $p_{rst} \ll 1$, which implies $e^{2(\beta_r + \beta_s + \beta_t)} \gg 1$. In this case,

$$p_{rst} \approx e^{-2(\beta_r + \beta_s + \beta_t)} \quad (26)$$

$$= e^{-2\beta_r} e^{-2\beta_s} e^{-2\beta_t} \quad (27)$$

Now, we still need to determine the elements of $\vec{\beta}$. To do this, note that

$$\langle k_i \rangle = \sum_s' \sum_t' p_{ist} \quad (28)$$

$$= \sum_s' \sum_t' e^{-2\beta_i} e^{-2\beta_s} e^{-2\beta_t} \quad (29)$$

$$= e^{-2\beta_i} \sum_s' \sum_t' e^{-2\beta_s} e^{-2\beta_t} \quad (30)$$

$$= e^{-2\beta_i} \sum_s' e^{-2\beta_s} \sum_t' e^{-2\beta_t} \quad (31)$$

$$\approx e^{-2\beta_i} \left(\sum_s' e^{-2\beta_s} \right)^2 \quad (32)$$

where the primes on the sums indicate that repeated indices are excluded. Now, assume that

$$\sum_s' e^{-2\beta_s} \approx \sum_s e^{-2\beta_s} = C \quad (33)$$

where C is a constant. Then,

$$e^{-2\beta_i} = \frac{\langle k_i \rangle}{C^2} \quad (34)$$

To determine C note that the average number of hyperedges is

$$\langle M \rangle = \sum_{i < j < k} p_{ijk} \quad (35)$$

$$= \frac{1}{3!} \sum_{i,j,k} p_{ijk} \quad (36)$$

$$\approx \frac{1}{3!} \left(\sum_i e^{-2\beta_i} \right)^3 \quad (37)$$

$$= \frac{1}{3!} C^3 \quad (38)$$

or

$$C = (3! \langle M \rangle)^{1/3} \quad (39)$$

Thus, finally, for sparse 3-regular simple hypergraphs

$$p_{rst} \approx \frac{\langle k_r \rangle \langle k_s \rangle \langle k_t \rangle}{(3! \langle M \rangle)^2} \quad (40)$$

Then, setting

$$\langle k_r \rangle = \bar{k}_r \quad (41)$$

the prescribed average degree, and

$$\langle M \rangle = \frac{1}{3} \sum_{i=1}^N \bar{k}_i \quad (42)$$

get

$$p_{rst} \approx \frac{\bar{k}_r \bar{k}_s \bar{k}_t}{\left(2 \sum_{i=1}^N \bar{k}_i\right)^2} \quad (43)$$

Hypergraphs generated with this hyperedge probability are the generalization of the Chung-Lu model for graphs.

2.2 g -regular Hypergraphs

It is easy to generalize the above result to g -regular hypergraphs. The results are that

$$\mathcal{H}(G; \vec{\beta}) = (g-1)! \sum_{i_1 < i_2 < \dots < i_g} \left(\sum_{q=1}^g \beta_{i_q} \right) a_{i_1 i_2 \dots i_g} \quad (44)$$

$$\mathcal{Z}(\vec{\beta}) = \prod_{i_1 < i_2 < \dots < i_g} \left(1 + e^{-(g-1)! \sum_{q=1}^g \beta_{i_q}} \right) \quad (45)$$

$$p_{i_1 i_2 \dots i_g} = \frac{1}{1 + e^{(g-1)! \sum_{q=1}^g \beta_{i_q}}} \quad (46)$$

In the sparse hypergraph limit

$$p_{i_1 i_2 \dots i_g} \approx e^{-(g-1)! \sum_{q=1}^g \beta_{i_q}} = \prod_{q=1}^g e^{-(g-1)! \beta_{i_q}} \quad (47)$$

$$e^{-(g-1)! \beta_i} = \frac{\langle k_i \rangle}{C^{g-1}} \quad (48)$$

$$\langle M \rangle \approx \frac{1}{g!} C^g \quad (49)$$

or

$$C \approx (g! \langle M \rangle)^{1/g} \quad (50)$$

Thus,

$$p_{i_1 i_2 \dots i_g} \approx \frac{\prod_{q=1}^g \langle k_{i_q} \rangle}{(g! \langle M \rangle)^{g-1}} \quad (51)$$

or

$$p_{i_1 i_2 \dots i_g} \approx \frac{\prod_{q=1}^g \bar{k}_{i_q}}{\left((g-1)! \sum_{j=1}^N \bar{k}_j\right)^{g-1}} \quad (52)$$

3 Prescribed Expected Joint Vertex Degrees

We wish to extend the earlier formulation to randomly generate g -regular hypergraphs with α -dimensional prescribed expected degrees

$$\bar{\mathbf{K}} \quad (53)$$

which is an $\alpha < g$ dimensional tensor, with elements

$$\bar{k}_{i_1, i_2, \dots, i_\alpha} \quad (54)$$

specifying the number of expected hyperedges incident on vertices $i_1, i_2, \dots, i_\alpha$. Assuming there are N vertices, there are N^α elements of $\bar{\mathbf{K}}$.

Calling the ensemble of all possible simple g -regular hypergraphs \mathcal{G}_g , the expected joint degrees are

$$\bar{k}_{i_1, i_2, \dots, i_\alpha} = \sum_{G \in \mathcal{G}_g} k_{i_1, i_2, \dots, i_\alpha}(G) P(G) \quad (55)$$

where $k_{i_1, i_2, \dots, i_\alpha}(G)$ is the joint degree of vertices $i_1, i_2, \dots, i_\alpha$ in a graph G and $P(G)$ is the probability of generating that graph.

Maximizing entropy while attempting to enforce the prescribed degree constraints, the probability of generating a hypergraph is

$$P(G; \mathbf{B}) = \frac{1}{\mathcal{Z}(\mathbf{B})} e^{-\mathbf{B} \cdot \mathbf{K}} = \frac{1}{\mathcal{Z}(\mathbf{B})} e^{-\sum_{i_1=1}^N \sum_{i_2=1}^N \dots \sum_{i_\alpha=1}^N \beta_{i_1, i_2, \dots, i_\alpha} k_{i_1, i_2, \dots, i_\alpha}} \quad (56)$$

where \mathbf{B} is an α -dimensional tensor of Laplace coefficients, and \mathbf{K} is the α -dimensional tensor of joint vertex degrees. Only elements of \mathbf{B} with nonrepeating elements are relevant, and all elements with permuted elements are equivalent.

Now, a g -regular hypergraph can be represented by a g -dimensional adjacency tensor \mathbf{a} , with elements of

$$a_{i_1 i_2 \dots i_g} = \begin{cases} 1 & \text{If a hyperedge exists connecting vertices } i_1, i_2, \dots, i_{g-1} \text{ and } i_g. \\ 0 & \text{Otherwise.} \end{cases} \quad (57)$$

Require that all elements that have permuted indices are equal, and all with any repeated elements are zero.

Note that

$$k_{i_1, i_2, \dots, i_\alpha} = \sum_{i_{\alpha+1}=1}^N \sum_{i_{\alpha+2}=1}^N \dots \sum_{i_g=1}^N a_{i_1 i_2 \dots i_g} \quad (58)$$

Recognizing Eqn. 56 as a Gibbs distribution, identify the ‘‘Hamiltonian’’

$$\mathcal{H}(G; \mathbf{B}) = \mathbf{B} \cdot \mathbf{K} \quad (59)$$

$$= \sum_{i_1=1}^N \sum_{i_2=1}^N \dots \sum_{i_\alpha=1}^N \beta_{i_1, i_2, \dots, i_\alpha} k_{i_1, i_2, \dots, i_\alpha} \quad (60)$$

$$= \sum_{i_1=1}^N \sum_{i_2=1}^N \dots \sum_{i_\alpha=1}^N \beta_{i_1, i_2, \dots, i_\alpha} \sum_{i_{\alpha+1}=1}^N \sum_{i_{\alpha+2}=1}^N \dots \sum_{i_g=1}^N a_{i_1 i_2 \dots i_g} \quad (61)$$

$$= \sum_{i_1=1}^N \sum_{i_2=1}^N \dots \sum_{i_g=1}^N \beta_{i_1, i_2, \dots, i_\alpha} a_{i_1 i_2 \dots i_g} \quad (62)$$

To learn how to proceed, consider the case with $g = 3$ and $\alpha = 2$, then

$$\mathcal{H}(G; \mathbf{B}) = \sum_{i=1}^N \sum_{j=1}^N \sum_{k=1}^N \beta_{ij} a_{ijk} \quad (63)$$

$$\begin{aligned} &= \sum_{i < j < k} \beta_{ij} a_{ijk} + \sum_{i < k < j} \beta_{ij} a_{ijk} + \sum_{j < i < k} \beta_{ij} a_{ijk} \\ &\quad + \sum_{j < k < i} \beta_{ij} a_{ijk} + \sum_{k < i < j} \beta_{ij} a_{ijk} + \sum_{k < j < i} \beta_{ij} a_{ijk} \end{aligned} \quad (64)$$

$$\begin{aligned} &= \sum_{i < j < k} \beta_{ij} a_{ijk} + \sum_{i < j < k} \beta_{ik} a_{ijk} + \sum_{i < j < k} \beta_{ji} a_{ijk} \\ &\quad + \sum_{i < j < k} \beta_{jk} a_{ijk} + \sum_{i < j < k} \beta_{ki} a_{ijk} + \sum_{i < j < k} \beta_{kj} a_{ijk} \end{aligned} \quad (65)$$

$$= 2 \left(\sum_{i < j < k} \beta_{ij} a_{ijk} + \sum_{i < j < k} \beta_{ik} a_{ijk} + \sum_{i < j < k} \beta_{jk} a_{ijk} \right) \quad (66)$$

$$= 2 \sum_{i < j < k} (\beta_{ij} + \beta_{ik} + \beta_{jk}) a_{ijk} \quad (67)$$

So then, for general g and α get

$$\mathcal{H}(G; \mathbf{B}) = \sum_{i_1=1}^N \sum_{i_2=1}^N \cdots \sum_{i_g=1}^N \beta_{i_1, i_2, \dots, i_g} a_{i_1 i_2 \dots i_g} \quad (68)$$

$$= (g - \alpha)! \alpha! \sum_{i_1 < i_2 < \dots < i_g} \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \cdots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right) a_{i_1 i_2 \dots i_g} \quad (69)$$

The “partition function” then is

$$\mathcal{Z}(\mathbf{B}) = \sum_{G \in \mathcal{G}_g} e^{-\mathcal{H}(G; \mathbf{B})} \quad (70)$$

$$= \prod_{i_1 < i_2 < \dots < i_g} \sum_{a_{i_1 i_2 \dots i_g} = 0}^1 e^{-(g-\alpha)! \alpha! \sum_{i_1 < i_2 < \dots < i_g} \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \cdots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right) a_{i_1 i_2 \dots i_g}} \quad (71)$$

$$= \prod_{i_1 < i_2 < \dots < i_g} \left(1 + e^{-(g-\alpha)! \alpha! \sum_{i_1 < i_2 < \dots < i_g} \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \cdots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right) a_{i_1 i_2 \dots i_g}} \right) \quad (72)$$

The probability that a hyperedge exists between vertices i_1, i_2, \dots , and i_g is (assume that

$$i_1 < i_2 < \dots < i_g)$$

$$p_{i_1 i_2 \dots i_g} = \langle a_{i_1 i_2 \dots i_g} \rangle \quad (73)$$

$$= \sum_{G \in \mathcal{G}_g} a_{i_1 i_2 \dots i_g}(G) P(G; \mathbf{B}) \quad (74)$$

$$= \frac{e^{-(g-\alpha)! \alpha! \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \dots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right)}}{1 + e^{-(g-\alpha)! \alpha! \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \dots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right)}} \quad (75)$$

$$= \frac{1}{1 + e^{(g-\alpha)! \alpha! \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \dots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right)}} \quad (76)$$

In the sparse hypergraph limit require $p_{i_1 i_2 \dots i_g} \ll 1$, which implies $e^{(g-\alpha)! \alpha! \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \dots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right)} \gg 1$. In this case,

$$p_{i_1 i_2 \dots i_g} \approx e^{-(g-\alpha)! \alpha! \left(\sum_{q_1=1}^g \sum_{q_2 > q_1}^g \dots \sum_{q_\alpha > q_{\alpha-1}}^g \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}} \right)} \quad (77)$$

$$= \prod_{q_1=1}^g \prod_{q_2 > q_1}^g \dots \prod_{q_\alpha > q_{\alpha-1}}^g e^{-(g-\alpha)! \alpha! \beta_{i_{q_1}, i_{q_2}, \dots, i_{q_\alpha}}} \quad (78)$$

Now, we still need to determine the elements of \mathbf{B} . To see how do this, consider again the case with $g = 3$ and $\alpha = 2$. Note that

$$p_{ijk} = e^{-2\beta_{ij}} e^{-2\beta_{ik}} e^{-2\beta_{jk}} \quad (79)$$

and that

$$\langle k_{ij} \rangle = \sum_s' p_{ijs} \quad (80)$$

$$= \sum_s' e^{-2\beta_{ij}} e^{-2\beta_{is}} e^{-2\beta_{js}} \quad (81)$$

$$= e^{-2\beta_{ij}} \sum_s' e^{-2\beta_{is}} e^{-2\beta_{js}} \quad (82)$$

$$\langle k_i \rangle = \frac{1}{2} \sum_s' \sum_t' p_{ist} \quad (83)$$

$$= \frac{1}{2} \sum_s' \sum_t' e^{-2\beta_{is}} e^{-2\beta_{it}} e^{-2\beta_{st}} \quad (84)$$

where the primes on the sums indicate that repeated indices are excluded. Also note that the average number of hyperedges is

$$\langle M \rangle = \sum_{i < j < k} p_{ijk} \quad (85)$$

$$= \frac{1}{3!} \sum_{i,j,k} p_{ijk} \quad (86)$$

$$\approx \frac{1}{3!} \sum_{i,j,k} e^{-2\beta_{ij}} e^{-2\beta_{ik}} e^{-2\beta_{jk}} \quad (87)$$

From Eq. 82 get

$$e^{-2\beta_{ij}} = \frac{\langle k_{ij} \rangle}{\sum_s' e^{-2\beta_{is}} e^{-2\beta_{js}}} \quad (88)$$

$$\approx \frac{\langle k_{ij} \rangle}{\sum_s e^{-2\beta_{is}} e^{-2\beta_{js}}} \quad (89)$$

$$= \frac{\langle k_{ij} \rangle}{C_{ij}} \quad (90)$$

where

$$C_{ij} \equiv \sum_s e^{-2\beta_{is}} e^{-2\beta_{js}} \quad (91)$$

Then also,

$$\langle k_i \rangle = \frac{1}{2} \sum_s' \sum_t' e^{-2\beta_{is}} e^{-2\beta_{it}} e^{-2\beta_{st}} \quad (92)$$

$$\approx \frac{1}{2} \sum_s \sum_t e^{-2\beta_{is}} e^{-2\beta_{it}} e^{-2\beta_{st}} \quad (93)$$

$$= \frac{1}{2} \sum_s e^{-2\beta_{is}} \sum_t e^{-2\beta_{it}} e^{-2\beta_{st}} \quad (94)$$

$$= \frac{1}{2} \sum_s e^{-2\beta_{is}} C_{is} \quad (95)$$

and

$$\langle M \rangle \approx \frac{1}{3!} \sum_{i,j,k} e^{-2\beta_{ij}} e^{-2\beta_{ik}} e^{-2\beta_{jk}} \quad (96)$$

$$= \frac{1}{3!} \sum_{i,j} e^{-2\beta_{ij}} \sum_k e^{-2\beta_{ik}} e^{-2\beta_{jk}} \quad (97)$$

$$= \frac{1}{3!} \sum_{i,j} e^{-2\beta_{ij}} C_{ij} \quad (98)$$

Conclusion: It seems like the best we can do is, using Eqs. 79, 89, and 91, to find the β_{ijs} by solving the simultaneous set of equations given by

$$e^{-2\beta_{ij}} = \frac{\bar{k}_{ij}}{\sum_s e^{-2\beta_{is}} e^{-2\beta_{js}}}, \quad \forall i, j \quad (99)$$

then use them to determine the C s in

$$p_{ijk} = \frac{\bar{k}_{ij} \bar{k}_{ik} \bar{k}_{jk}}{C_{ij} C_{ik} C_{jk}} \quad (100)$$

On the other hand, one could work directly with the C s instead of the β s. Note that

$$e^{-2\beta_{ij}} \approx \frac{\bar{k}_{ij}}{C_{ij}} \quad (101)$$

$$\sum_j e^{-2\beta_{ij}} e^{-2\beta_{jk}} \approx \sum_j \frac{\bar{k}_{ij} \bar{k}_{jk}}{C_{ij} C_{jk}} \quad (102)$$

$$C_{ik} \approx \sum_j \frac{\bar{k}_{ij} \bar{k}_{jk}}{C_{ij} C_{jk}} \quad \forall i, k \quad (103)$$

So that one can solve the system of simultaneous equations given by Eqns. 103 to determine the C s, and then use them in Eq. 100 to generate the hypergraph.