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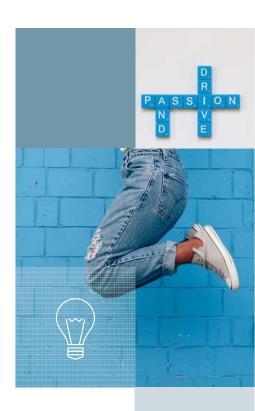
The definition of hypergraph

(1)

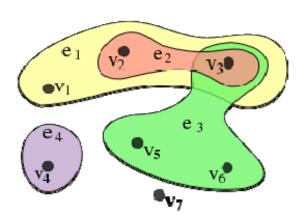
- In a graph, an edge, only two vertices
- In a hypergraph, an edge, more than two vertices.

(2)

- A hypergraph HG=(V,E)
- V : vertex set.
- V : vertex set.
- e ∈ E is a subset of V.



An example of hypergraph



An example of a hypergraph

$$V = \{v_1, v_2, v_3, v_4, v_5, v_6, v_7\}$$

$$E = \{e_1, e_2, e_3, e_4\}$$

=
$$\{\{v_1, v_2, v_3\}, \{v_2, v_3\},$$

$$\{v_3, v_5, v_6\}, \{v_4\}\}.$$



The application of hypergraph in data mining

Discovering semantic associations between items

Clustering

Spatial data mining



Why we use hypergraph

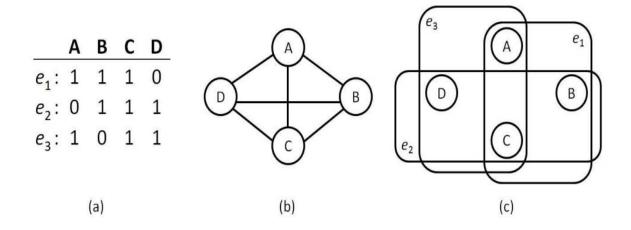
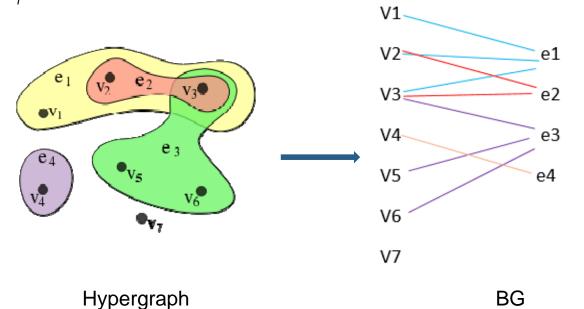


Figure 1. (a) an example transaction table; (b) the Gaifman graph representation of the table; (c) The hypergraph representation of the table

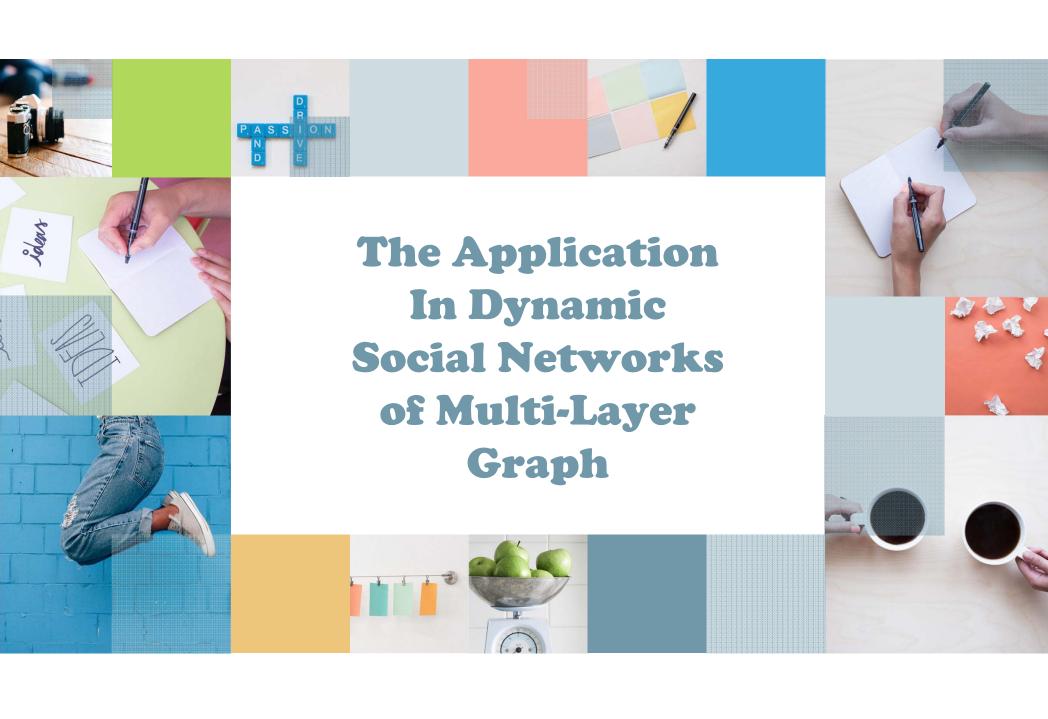


How to transform hypergraph to bipartite graph(BG)

- ➤ Let V and E be the two parts of BG
- (v_1, e_1) are connected with an edge if and only if vertex v_1 is contained in edge e_1 in H







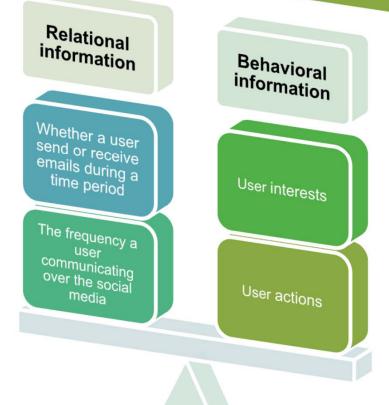
We are married



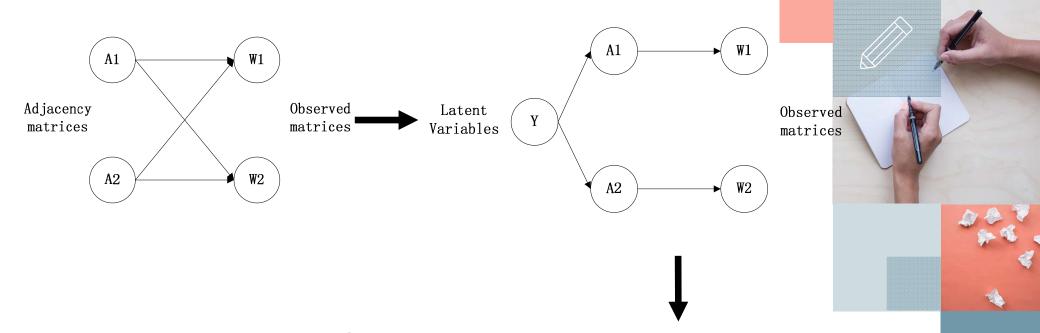




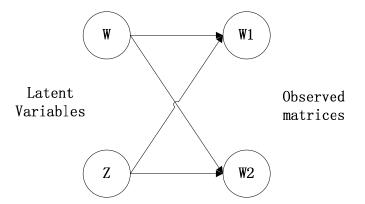
Two type of sources in the connectivity information



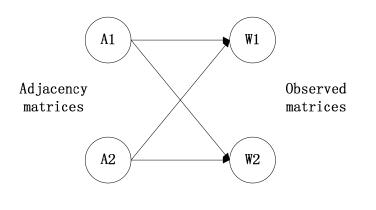




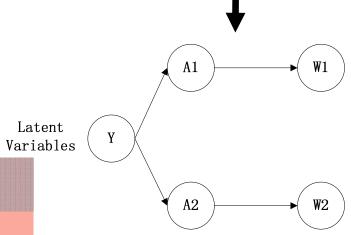
The evolution of solutions of the multi-layer graph



Hierarchical Latent-variable Model



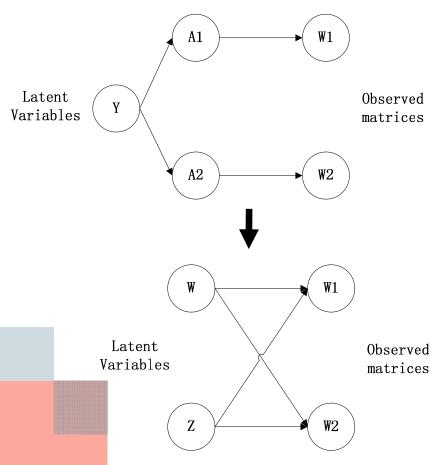




Observed matrices

$$P(Y|W_1, W_2) = \sum_{A_1, A_2} P(Y|A_1, A_2) P(A_1, A_2|W_1, W_2).$$

Posterior Mixture Model



$$P(Y|W_1, W_2) = \sum_{A_1, A_2} P(Y|A_1, A_2) P(A_1, A_2|W_1, W_2).$$

$$P(W|W_1, W_2)$$

$$= \xi \frac{P(W)P_1(W_1|W)}{P_1(W_1)} + (1 - \xi) \frac{P(W)P_2(W_2|W)}{P_2(W_2)}$$

$$= P(W) \left[\gamma_1 P_1(W_1|W) + \gamma_2 P_2(W_2|W) \right],$$
(13)
$$= P(W) \left[\gamma_1 P_1(W_1|W) + \gamma_2 P_2(W_2|W) \right],$$
(14)

$$\arg\max_{W} \left[\gamma_1 P_1(W_1|W) + \gamma_2 P_2(W_2|W) \right].$$
 (15)

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Thank You

